

## **An Evaluation of ENSO Asymmetry in the Community Climate System Models: A View from the Subsurface**

TAO ZHANG AND DE-ZHENG SUN

*Cooperative Institute for Research in Environmental Sciences, University of Colorado, and NOAA/Earth System Research Laboratory/Physical Sciences Division, Boulder, Colorado*

RICHARD NEALE

*National Center for Atmospheric Research, Boulder, Colorado*

PHILIP J. RASCH

*Pacific Northwest National Laboratory, Richland, Washington*

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### ABSTRACT

The asymmetry between El Niño and La Niña is a key aspect of ENSO that needs to be simulated well by models in order to fully capture the role of ENSO in the climate system. Here the asymmetry between the two phases of ENSO in five successive versions of the Community Climate System Model (CCSM1, CCSM2, CCSM3 at T42 resolution, CCSM3 at T85 resolution, and the latest CCSM3 + NR, with the Neale and Richter convection scheme) is evaluated. Different from the previous studies, not only is the surface signature of ENSO asymmetry examined, but so too is its subsurface signature. By comparing the differences among these models as well as the differences between the models and the observations, an understanding of the causes of the ENSO asymmetry is sought.

An underestimate of the ENSO asymmetry is noted in all of the models, but the latest version with the Neale and Richter scheme (CCSM3 + NR) is getting closer to the observations than the earlier versions. The net surface heat flux is found to damp the asymmetry in the SST field in both the models and observations, but the damping effect in the models is weaker than that in the observations, thus excluding a role of the surface heat flux in contributing to the weaker asymmetry in the SST anomalies associated with ENSO. Examining the subsurface signatures of ENSO—the thermocline depth and the associated subsurface temperature for the western and eastern Pacific—reveals the same bias; that is, the asymmetry in the models is weaker than that in the observations.

The analysis of the corresponding Atmospheric Model Intercomparison Project (AMIP) runs in conjunction with the coupled runs suggests that the weaker asymmetry in the subsurface signatures in the models is related to the lack of asymmetry in the zonal wind stress over the central Pacific, which in turn is due to a lack of sufficient asymmetry in deep convection (i.e., the nonlinear dependence of the deep convection on SST). In particular, the lack of a westward shift in the deep convection in the models in response to a cold phase SST anomaly is found as a common factor that is responsible for the weak asymmetry in the models. It is also suggested that a more eastward extension of the deep convection in response to a warm phase SST anomaly may also help to increase the asymmetry of ENSO. The better performance of CCSM3 + NR is apparently linked to an enhanced convection over the eastern Pacific during the warm phase of ENSO. Apparently, either a westward shift of deep convection in response to a cold phase SST anomaly or an increase of convection over the eastern Pacific in response to a warm phase SST anomaly leads to an increase in the asymmetry of zonal wind stress and therefore an increase in the asymmetry of subsurface signal, favoring an increase in ENSO asymmetry.

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*Corresponding author address:* Dr. Tao Zhang, 325 Broadway, R/PSD1, CIRES and NOAA/PSD, Boulder, CO 80305.  
E-mail: tao.zhang@noaa.gov

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## 1. Introduction

El Niño–Southern Oscillation (ENSO) is one of the most important natural modes of the climate system and has been extensively studied (Philander 1990; Sun 2007). Two new dimensions of ENSO have emerged from the more recent investigation: the nonlinear and diabatic aspects of ENSO (Sun 2003; Sun et al. 2004; Sun and Zhang 2006). The sensitivity of ENSO to diabatic heating and its role as a basin-scale heat mixer in the tropical Pacific are linked to a critical quantity of ENSO—the asymmetry between its two phases or, as referred in some other studies, the residual of ENSO (Rodgers et al. 2004; Schopf and Burgman 2006). Related, the ENSO asymmetry is potentially a mechanism for decadal variability (Rodgers et al. 2004; Sun and Yu 2009) and a cause of the bias in the time-mean background state (Sun and Zhang 2006; Schopf and Burgman 2006). The asymmetry between the two phases of ENSO was first noted in the SST fields, which manifest as the maximum warming during the warm phase exceeding the maximum cooling during the cold phase [or a positive skewness in the probability density distribution of the Niño-3 SST anomaly (Burgers and Stephenson 1999)]. Later, the asymmetry has also been noted in the depth of the equatorial thermocline (Rodgers et al. 2004), in the upper-ocean heat content (Tang and Hsieh 2003), and in the subsurface temperature (Schopf and Burgman 2006; Sun and Zhang 2006).

The suggested physics mechanisms responsible for the asymmetry between El Niño and La Niña include the nonlinearity (or asymmetry) in the net surface heating or the upwelling cooling (Meinen and McPhaden 2000), the asymmetry of wind stress in the western and central tropical Pacific (Kang and Kug 2002), the nonlinear heating term in the surface heat budget equation (Jin et al. 2003), and a kinematic effect of oscillating a nonlinear temperature profile with finite-amplitude excursions that will cause the Eulerian time-mean temperature to rise (fall) where the curvature of the temperature is positive (negative) as the amplitude of the oscillations increases (Schopf and Burgman 2006). In assessing the relationship between ENSO SST anomalies and the warm water volume (WWV), Meinen and McPhaden (2000) noted that for the same WWV anomaly, the corresponding warm El Niño SST anomaly is larger than the cold La Niña SST anomaly, and hence suggested that the relative importance of physical processes that control the growth of the warm and cold SST anomalies must be different. Specifically, they mentioned that the ENSO asymmetry may be caused by the asymmetry in the net surface heat flux into the ocean or the upwelling cooling. Although the net surface heat flux acts as a negative feedback that damps the growth of the ENSO SST

anomaly in both phases, it may be more effective at heating the ocean during cold phases of ENSO than it is at cooling the ocean during warm phases of ENSO. Alternately, the ability of upwelling and vertical mixing to cool the surface may saturate at some threshold beyond which further thermocline shoaling does not lead to further SST cooling. By the use of National Centers for Environmental Prediction reanalysis data and various ocean and atmosphere models, Kang and Kug (2002) suggested that the relatively weak SST anomalies during La Niña compared with those of El Niño are related to a westward shift of wind stress anomalies. They further attribute the westward shift of the wind stress anomalies to the nonlinear dependence of deep convection on the SST, which was noted earlier in Gadgil et al. (1984) and Hoerling et al. (1997). Jin et al. (2003) have suggested that nonlinear dynamical heating—the nonlinear term in the surface heat budget equation—may be a cause of the ENSO asymmetry. More recently, Schopf and Burgman (2006) demonstrated that oscillating a nonlinear temperature profile with finite-amplitude excursions can cause the Eulerian time-mean temperature to rise (fall) where the curvature of the temperature is positive (negative) as the amplitude of the oscillations increases. This mechanism is found to be able to mimic observed changes in the mean sea surface temperatures in the Pacific between the 1920s, 1960s, and 1990s resulting from the changing ENSO amplitude.

After noting the positive skewness in the distribution of observed Niño-3 SST anomalies, Burgers and Stephenson (1999) evaluated the ENSO asymmetry in the early National Center for Atmospheric Research (NCAR) Climate System Model (CSM1.0; Boville and Gent 1998) and found that the model underestimates the asymmetry in the Niño-3 SST distribution. This bias in NCAR CSM1.0 was shown again by later studies that participated in the El Niño Simulation Intercomparison Project (ENSIP; Hannachi et al. 2003) and the Coupled Model Intercomparison Project (CMIP; An et al. 2005). Because the NCAR Community Climate System Model, version 1 (CCSM1; CSM1.0) severely underestimates the amplitude of ENSO variability, Burgers and Stephenson (1999) suggested a link between the bias in the asymmetry and the bias in the amplitude. The simulation of the amplitude of ENSO variability, measured by the variance of Niño-3 SST, in the late versions of the NCAR CCSM is greatly improved (Deser et al. 2006). Does the improved simulation of the amplitude of ENSO variability also lead to a better simulation in the asymmetry of ENSO?

To answer this question, and more generally to explore the causes of the ENSO asymmetry, we will evaluate the ENSO asymmetry in five successive versions of the NCAR coupled models—NCAR CCSM1 (Boville and

Gent 1998), NCAR CCSM, version 2 (CCSM2; Kiehl and Gent 2004), NCAR CCSM, version 3 (CCSM3) at T42 resolution, NCAR CCSM3 at T85 resolution (Collins et al. 2006a), and NCAR CCSM3 + NR (Neale et al. 2008). We will see that an improvement in simulating the variance of Niño-3 SST does not guarantee an improvement in simulating the skewness.

Recognizing that the El Niño warming and La Niña cooling are driven by the vertical advection associated with subsurface temperature changes (Zebiak and Cane 1987), we will examine the asymmetry in the subsurface temperature. By examining the subsurface signal, we reasonably hope that we may obtain more information about the causes of bias in the simulated ENSO in CCSM, particularly in view of the findings by Sun and Zhang (2006) and Sun (2007). In these studies, they found that ENSO has a time-mean effect on the vertical temperature structure of the upper ocean; ENSO acts as a vertical heat mixer in the tropical Pacific. This time-mean effect is apparently linked to the asymmetry between the upper-ocean temperature anomalies during the two phases of ENSO. We will investigate the relationship between SST, convection, wind stress, and surface heat flux, particularly the asymmetry in these fields, attempting to discern and understand the mechanisms responsible for the asymmetry between the two phases of ENSO, and hence improve the processes involved that enable ENSO to have a “heat mixer” time-mean effect on the equatorial upper-ocean temperature.

This paper is organized as follows. In section 2, we introduce the methodology of our analysis and the models and observations used in the analysis. We present the assessment and analysis of ENSO asymmetry in the coupled NCAR models in section 3. We will first assess the ENSO asymmetry in the SST (section 3a) and then look at the asymmetry in upper-ocean temperature in the models (section 3b). We further analyze the corresponding fields of surface heat flux, precipitation, and surface wind stress (section 3c). One thing that emerges from the analysis of these fields from the coupled models is an association between a weak asymmetry in the zonal wind stress and a weak asymmetry in the SST anomalies in the models. We present an analysis of the relationship between wind stress and SST in the corresponding Atmospheric Model Intercomparison Project (AMIP) runs of these coupled models (section 3c) to demonstrate that the weak asymmetry in the zonal wind stress in the models is due to a less nonlinear dependence of deep convection on the underlying SST in the models, and thereby underscore the role of this nonlinearity in giving rise to the asymmetry of ENSO. A summary and discussion are given in section 4.

## 2. Methodology, data, and model

To quantify the asymmetry of ENSO, we will examine the skewness of the interannual variability of SST as well as the subsurface temperature of the equatorial upper Pacific. The skewness  $\sqrt{b_1}$  is defined as  $\sqrt{b_1} = m_3/(m_2)^{3/2}$ , where  $m_k$  is the  $k$ th moment,  $m_k = (\sum_{i=1}^N (X_i - \bar{X})^k)/N$ , and where  $X_i$  is the  $i$ th sample of variable  $X$ ,  $\bar{X}$  is the mean, and  $N$  is the number of samples (Burgers and Stephenson 1999). In addition to using skewness to measure the asymmetry of ENSO, we will do composites of El Niño and La Niña, and then use the sum of the composite of these two phases of ENSO to measure the asymmetry.

Sustained climate modeling work through NCAR and its interaction with the climate community in the United States and the world at large has made steady progress in building better and better climate system models. The effort has also created a suite of models with progressive changes, and offered a potential resource to address fundamental climate questions. Up until the writing of this paper, five major versions of NCAR models have been made available to the public. The five NCAR coupled models are NCAR CCSM1 (Boville and Gent 1998), NCAR CCSM2 (Kiehl and Gent 2004), NCAR CCSM3 at T42, NCAR CCSM3 at T85 (Collins et al. 2006a), and NCAR CCSM3 + NR (Neale et al. 2008). We will examine the ENSO asymmetry in all of these five versions and highlight the differences among these models and the progress made in simulating this particular aspect of ENSO.

The atmospheric component of CCSM1, CCSM2, CCSM3 at T42, and CCSM3 at T85 is the Community Climate Model, version 3 [CCM3; or Community Atmosphere Model (CAM1; Kiehl et al. 1998)], CAM, version 2 (CAM2; Kiehl and Gent 2004), CAM version 3 (CAM3) at T42 resolution, and CAM3 at T85 resolution (Collins et al. 2006b; Hack et al. 2006), respectively. From CCM3 (CAM1) to CAM2, there are several changes in physical parameterizations. First, the evaporation of convective precipitation back to the atmosphere was added in order to reduce the dry bias in the free troposphere in the model following Sundqvist (1988). Collins et al. (2002) also updated the longwave radiation treatment of water vapor to enhance the longwave cooling in the upper troposphere. In addition, the Rasch and Kristjansson (1998) treatment of stratiform clouds was modified according to Zhang et al. (2003), which predicts the total (liquid + ice) cloud condensate (Boville et al. 2006; Zhang and Sun 2006). From CAM2 to CAM3, the physics of cloud and precipitation process has been modified extensively by Boville et al. (2006). The modifications include (i) separate prognostic treatments of liquid and ice condensate,

(ii) advection, detrainment, and sedimentation of cloud condensate, (iii) separate treatments of frozen and liquid precipitation, and (iv) the calculation of the cloud fraction. The radiation code has been updated with new parameterizations for the longwave and shortwave interactions with water vapor (Collins et al. 2006a; Zhang and Sun 2006).

The ocean component of CSM1 is the NCAR CSM Ocean Model (NCOM), which is described in Gent et al. (1998). NCOM includes the mesoscale eddy parameterization of Gent and McWilliams (1990), and the vertical mixing scheme based on the  $K$ -profile parameterization scheme of Large et al. (1994). The ocean component of CCSM2 uses the Parallel Ocean Program (POP) code developed at the Los Alamos National Laboratory (Smith et al. 1992). The mesoscale eddy parameterization and the vertical mixing scheme are the same as those used in the ocean component of CSM1. The ocean component of CCSM3 is based upon the POP, version 1.4.3 (Smith and Gent 2002). The CCSM3 ocean model has two important improvements over CCSM2 as discussed by Danabasoglu et al. (2006). These are modifications to the Large et al. (1994)  $K$ -profile parameterization and implementation of spatially varying monthly solar absorption based on ocean color observations.

CCM3 (CAM1), CAM2, and CAM3 share the same convection scheme, however. The deep convection is parameterized following Zhang and McFarlane (1995). Shallow convection is parameterized according to Hack (1994).

CCSM3 + NR is a developing version based on its immediate predecessor, CCSM3 (Neale et al. 2008). It uses a modified deep convection scheme; the Neale and Richter convection scheme replaces the Zhang and McFarlane deep convection scheme (Zhang and McFarlane 1995) used in CCSM3. The changes to the existing parameterization of deep convection (Zhang and McFarlane 1995) are the inclusion of convective momentum transports (CMT) and a dilution approximation for the calculation of convective available potential energy (CAPE; Neale et al. 2008).

The observational SST data for our evaluation are from the Hadley Centre Sea Ice and SST (HadISST) dataset (Rayner et al. 2003). The surface heat flux data are from the 40-yr European Centre for Medium-Range Weather Forecasts (ECMWF) Re-Analyses (ERA-40; Uppala et al. 2005). The wind stress data are obtained from the simple ocean data assimilation (SODA) set (Carton and Giese 2008) in which the surface winds are a combination of ERA-40 and Quick Scatterometer (QuikSCAT) satellite observations. Observations of precipitation are obtained from the Climate Prediction Center (CPC) Merged Analysis of Precipitation (CMAP; Xie and Arkin

1997). The simple ocean data assimilation (SODA) set (Carton et al. 2000) is used for validating the upper-ocean temperature in the models.

### 3. Results

#### *a. Asymmetry in the SST fields*

Figure 1 shows histograms of the distribution of Niño-3 SST anomalies for observations (1950–99) and five NCAR coupled models. The histogram from the observations has a longer tail on the right. The maximum positive anomaly in the observations is close to 3.5°C while the maximum negative anomaly is limited to –2.5°C. This skewness toward warm El Niño events has been noted by previous studies (e.g., Burgers and Stephenson 1999; Trenberth 1997). The models, however, clearly have a more Gaussian-like distribution. There are progressive increases in the range of variability from CCSM1 to CCSM3 + NR. The range of the variability of Niño-3 SST in the latest version has in fact become very comparable to that of the observations. The frequency distribution of the anomalies in the latest version is more spread out within the range confined by the maximum positive and negative anomalies, compared to that in four old versions. In other words, there are relatively few instances with weak anomalies in the latest version. The overall distribution for Niño-3 anomalies is much closer to normal distribution in the NCAR models than in the observations. A positive skewness is evident in the histogram for the latest version, however.

Table 1 lists the skewness of Niño-3 SST anomalies from observations and the models, together with their variance. The observed skewness value of 0.86 is the same as that obtained by Burgers and Stephenson (1999). The standard deviation in the present study (0.88°C) is slightly smaller than the findings of Burgers and Stephenson (1999) (0.91°C). The possible reason is that HadISST data may underestimate the SST variability before 1981 because of the EOF filters used for the interpolation. The numbers in the table confirm the impression from the histograms that the models underestimate the skewness of Niño-3 SST anomalies. Among the five models, the latest (CCSM3 + NR) has the strongest ENSO variability and a significant positive skewness. It is interesting to note that there is a much weaker ENSO variability in CCSM1, but the skewness in CCSM1 is comparable to that in CCSM3 + NR. On the other hand, the variance of ENSO in CCSM2 and CCSM3 is quite comparable to that observed, but these two models almost have not skewness at all. Thus, an increase in the variance of Niño-3 SST anomalies does not necessarily lead to an increase in the skewness. A

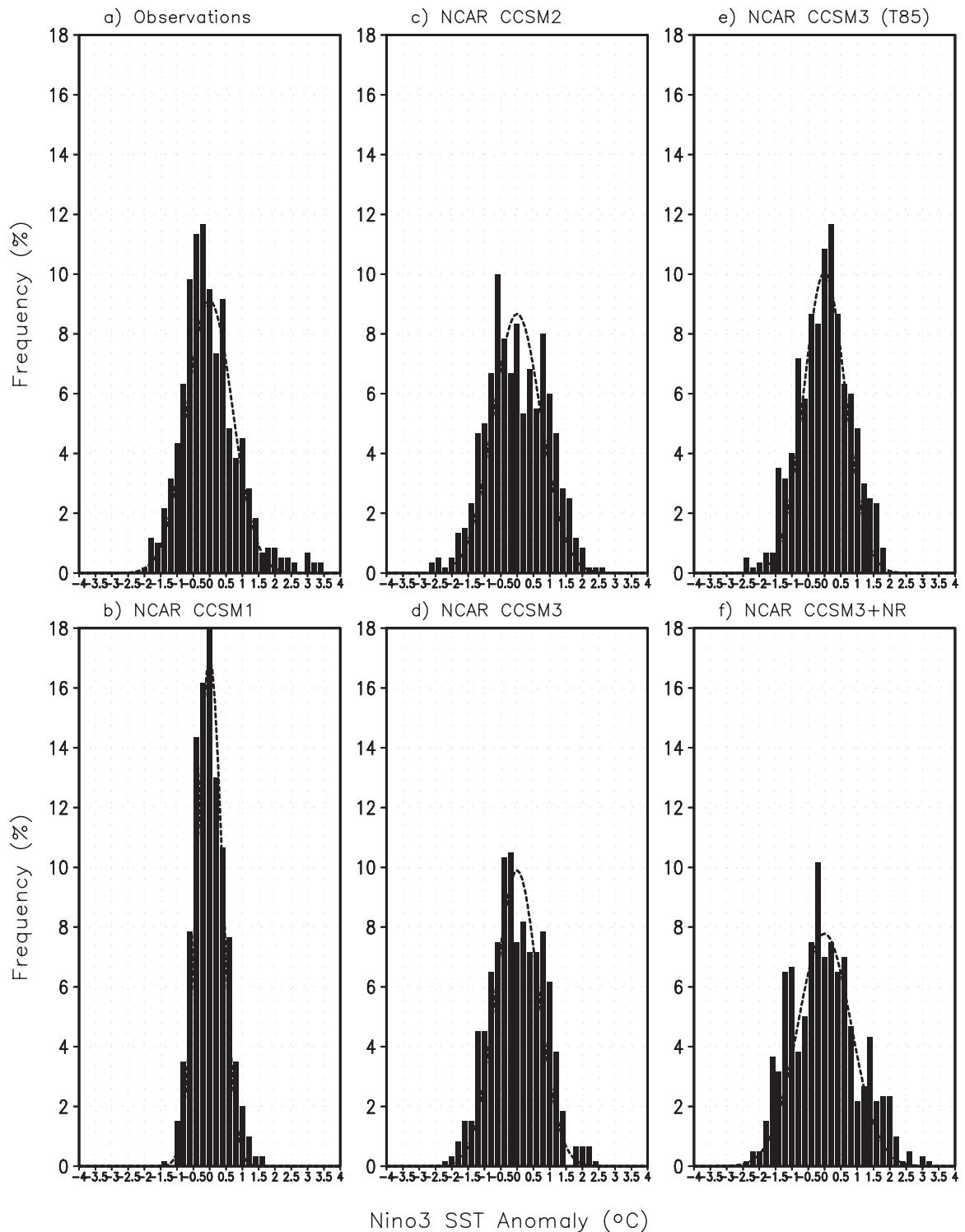


FIG. 1. Histograms of the monthly-mean Niño-3 SST anomalies from observations and five NCAR coupled models. The normal distributions with the respective standard deviation are plotted as dashed lines. A bin width of  $0.2^{\circ}\text{C}$  is used. The length of data used in the calculation is 50 yr for all the models and observations (1950–99).

TABLE 1. Statistics estimated for monthly Niño-3 time series from observations and five NCAR coupled models. The length of data for computing the statistics is 50 yr for all the models and observations (1950–99). The values listed in parentheses are the results from the 100-yr-long model data and observations (1900–99).

Source	Standard deviation	Skewness
Observed	0.88°C (0.82)	0.86 (0.62)
CCSM1	0.47°C (0.47)	0.36 (0.31)
CCSM2	0.92°C (0.92)	0.012 (0.052)
CCSM3	0.81°C (0.84)	0.089 (−0.10)
CCSM3 (T85)	0.79°C (0.85)	−0.13 (−0.20)
CCSM3 + NR	1.03°C (1.12)	0.31 (0.17)

related question raised by the results in Table 1 is that the variance alone may not be a good indicator for the actual level of ENSO activity in the models. For example, measured by the variance of Niño-3 SST anomalies alone, one would conclude that the level of ENSO variability in CCSM2, CCSM3, and CCSM3 at T85 is comparable to that of the observations, but quantifying the skewness of Niño-3 SST anomalies in these models indicates otherwise.

To get another measure of ENSO asymmetry, we construct composites of El Niño and La Niña events. Figure 2 gives the spatial pattern of composite SST anomalies during El Niño and La Niña events from observations and five NCAR models. In the observations, the positive SST anomalies associated with warm events tend to extend more to the South American coast compared to the negative anomalies associated with cold events. This feature is not evident in the models. Both the El Niño and La Niña composites in the observations have a wider meridional extension of the SST anomalies than in four earlier versions of NCAR models (CCSM1, CCSM2, and the two CCSM3 models). Positive SST anomalies associated with El Niño and negative SST anomalies associated with La Niña in these models also extend too far westward, consistent with the existence of an excessive cold tongue in these models (Sun et al. 2006). The latest CCSM3 + NR has an improvement in the zonal and meridional extension of the SST anomalies, and this has been noted already by Neale et al. (2008). The observed maximum SST anomaly of 1.6°C around 110°W is well captured in CCSM3 + NR during the warm phase, while the negative SST anomaly is somewhat overestimated in the model during the cold phase.

The spatial distribution of the asymmetry between El Niño and La Niña events is shown in Fig. 3. The asymmetry pattern is defined as the sum of El Niño and La Niña composites. In the observations, the asymmetry map has a similar spatial structure to that of the El Niño

events (Fig. 3a). The figure is characterized by positive SST anomalies of about 0.2°–0.5°C over the central and eastern tropical Pacific and small negative anomalies of about −0.1°C over the western tropical Pacific. The main bias in the models is a severe underestimation of positive anomalies over the eastern tropical Pacific (Figs. 3b–f). The skewness pattern (right panel contour) is very similar to the pattern of the sum of warm and cold events (left panel) and indicates again an underestimate of positive skewness over the eastern tropical Pacific in the models. It appears that there is hardly any SST asymmetry in the four previous versions. By comparison, the latest CCSM3 + NR shows a significant improvement, but the skewness in the model does not even reach half of the observed value in the eastern Pacific, and has a stronger negative skewness in the western Pacific. We have also noted that over the central Pacific, the cold bias in SST climatology (shaded) in the NCAR CCSM1 and CCSM2 is accompanied with a positive skewness opposite to that observed (Figs. 3h,i). As remarked by Meinen and McPhaden (2000), in the upwelling zone the depth of the thermocline is more bounded on its lower side than its upper side—the depth can be shallower if it already reaches the surface. This asymmetry would favor a positive skewness in the SST anomaly. This reasoning also suggests that a cold bias in the mean SST in the central Pacific resulting from a shallower thermocline should contribute to a positive bias in the skewness, which is apparently what we see here. The warm bias in SST climatology (shaded) in the western Pacific, on the other hand, is accompanied with a stronger negative skewness (Figs. 3j,l). This is probably due to the sensitivity of deep convection to the total SST, because the upwelling is very weak over these regions. Deep convection tends to occur only when the total SST is sufficiently high. Through the albedo of its anvils and the efficient horizontal heat dispersion in the tropics, deep convection has been shown to damp the growth of a warm SST anomaly (Ramanathan and Collins 1991; Wallace 1992; Sun and Trenberth 1998).

#### b. Asymmetry in the subsurface temperature

Next we will look at the asymmetry of the subsurface signal. As demonstrated in the pioneering studies of ENSO (Zebiak and Cane 1987, and others), the SST anomaly in the eastern Pacific associated ENSO is driven by the subsurface temperature anomaly in that region, which is in turn related to the subsurface temperature anomaly in the western Pacific. Figure 4 shows the equatorial upper-ocean temperature anomalies during both phases of ENSO. Observations show a maximum positive anomaly of 2.5°C during the warm phase in the eastern Pacific (Fig. 4a) and a maximum negative

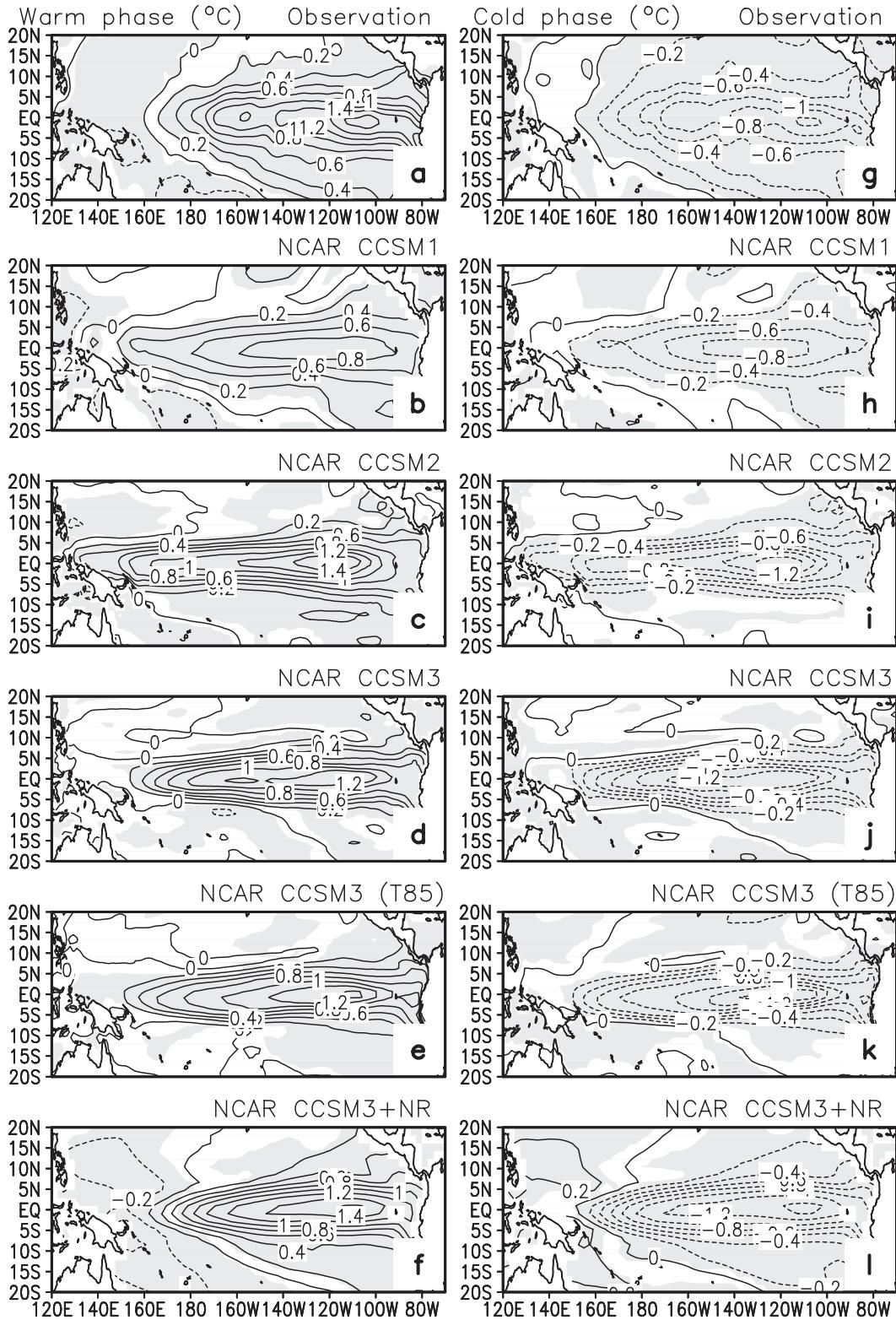


FIG. 2. Spatial distributions of composite SST anomalies during (left) El Niño and (right) La Niña events from observations and five NCAR coupled models. The positive (negative) anomalies of Niño-3 SST with a value greater than  $0.5^{\circ}\text{C}$  ( $-0.5^{\circ}\text{C}$ ) are selected to construct composites of warm (cold) events. The length of data used in the calculation is 50 yr for all of the NCAR models and observations (1950–99). Shading indicates that the anomalies are statistically significant at the 95% confidence level using the  $t$  test.

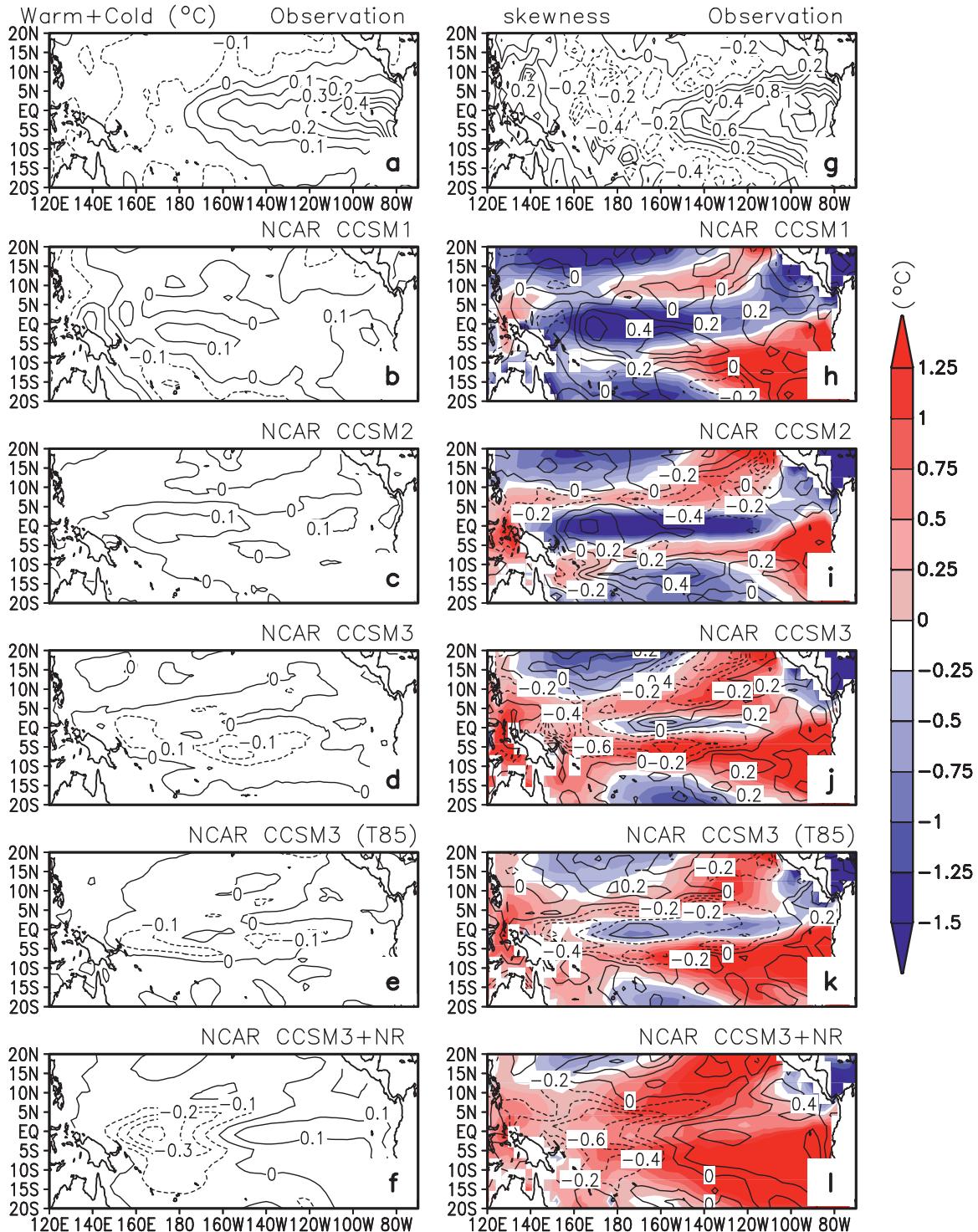


FIG. 3. Spatial distributions of the asymmetry between El Niño and La Niña events from observations and five NCAR coupled models. (left) The sum of El Niño and La Niña composite SST anomalies, and (right) the skewness pattern of the SST anomalies (contour) overlaid with the difference in the SST climatology between the models and observations (shaded). The positive (negative) anomalies of Niño-3 SST with a value greater than  $0.5^{\circ}\text{C}$  ( $-0.5^{\circ}\text{C}$ ) are selected to construct composites of warm (cold) events. The length of data used in the calculation is 50 yr for all of the models and observations (1950–99).

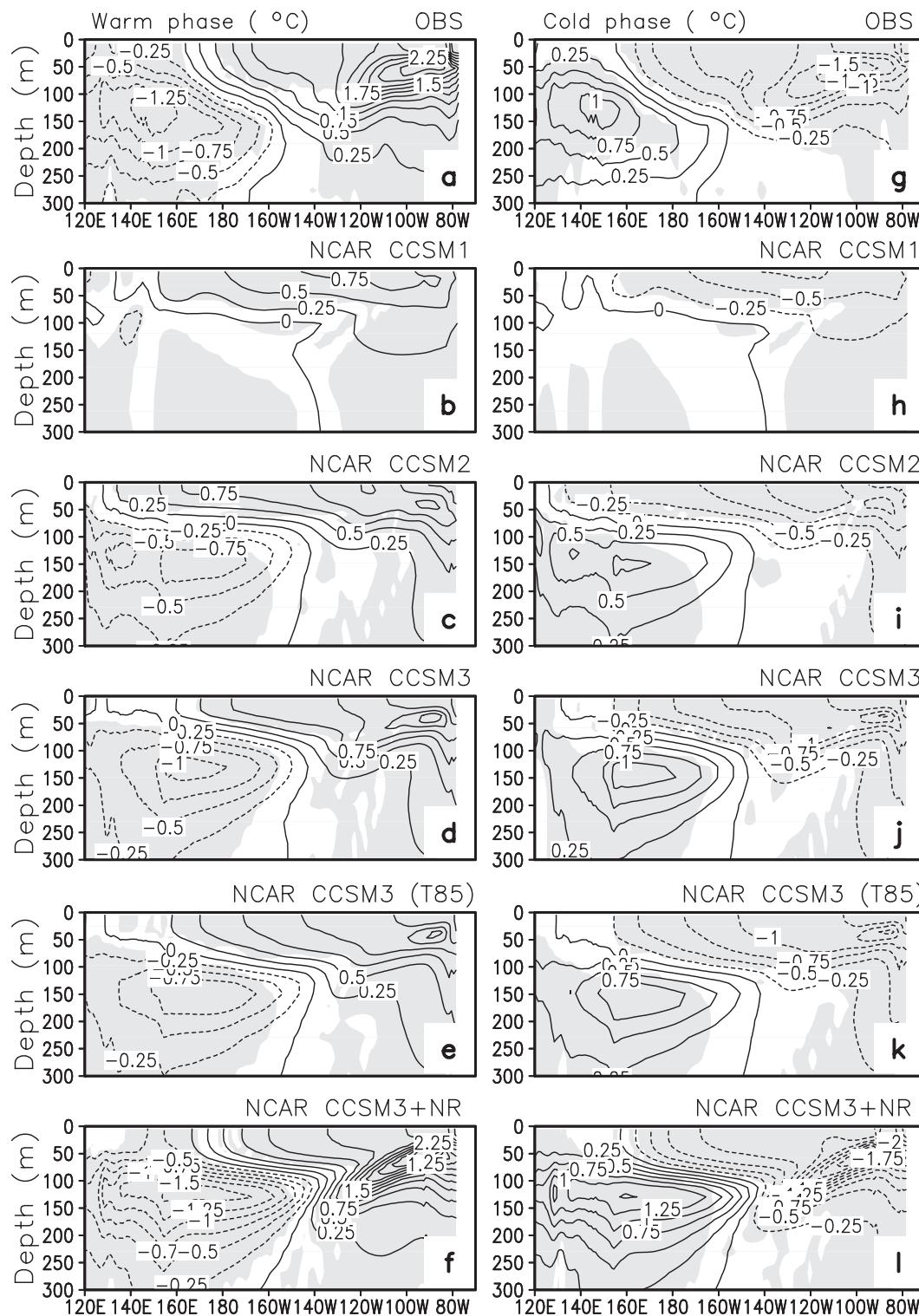


FIG. 4. The composite anomalies of equatorial ( $5^{\circ}\text{S}$ – $5^{\circ}\text{N}$ ) upper-ocean temperature during (left) warm and (right) cold periods. The length of data used in the calculation is 50 yr for all the five models and SODA data (1950–99). Shading indicates that the anomalies are statistically significant at the 95% confidence level using the  $t$  test.

anomaly of  $-1.5^{\circ}\text{C}$  during the cold phase in the eastern Pacific (Fig. 4g). The location of the maximum response occurs approximately in the region of  $120^{\circ}\text{W}$  to the eastern coast at a depth of about 50–100 m. During the warm phase, four old versions of CCSM have a much weaker positive subsurface temperature anomaly over the eastern Pacific. The underestimate of the anomaly ranges from about  $2^{\circ}\text{C}$  in CCSM1 to  $1^{\circ}\text{C}$  in CCSM3. The positive anomalies in the models are more confined to the surface and extend too far to the west. These anomalies in the mixed layer appear to be caused by processes associated with changes in upwelling, heat flux, and vertical mixing over most regions. The accompanying negative temperature anomalies in the subsurface of the western Pacific are also underestimated in these models. The maximum negative temperature anomaly occurs at about the same depth as that in the observations, but is more eastward. The axis of maximum negative temperature anomaly in the depth–longitude plane is more horizontal in the models than in the observations. In terms of magnitude, the upper-ocean temperature anomalies in the latest version, CCSM3 + NR, are comparable to those in the observations.

During the cold phase, the models do a better job in simulating the upper-ocean temperature anomalies than in the warm phase. The negative temperature anomalies in CCSM1 and CCSM2 are underestimated, but those in the two versions of CCSM3 are comparable to that observed over the eastern Pacific. The latest version of CCSM3 + NR has stronger subsurface temperature anomalies than those in the observations. CCSM3 + NR differs from the observed value in its simulation of the subsurface temperature anomaly over the western Pacific by as much as  $0.25^{\circ}\text{C}$ . Again, we noted that over the western Pacific the shape of subsurface temperature anomalies in observations differs from those in the models—the former have a more downward tilt from the west to the east while the latter are more horizontal.

Figure 5 shows the time-mean thermocline depth (indicated by a  $20^{\circ}\text{C}$  isotherm depth) in the observations and models. The observed thermocline depth over the western Pacific increases from 150 to around 175 m, showing a downward tilt from the west to the east. The modeled thermocline depth is either flat or has a slightly upward tilt from the west to the east in the western Pacific. In terms of zonal extent, there is a progressive increase in the size of the western Pacific warm pool.

Figure 6 shows a basinwide view of the thermocline depth anomalies during the warm and cold phases. All of the models except CCSM1 can simulate the observed feature: the thermocline depth is increased in the tropical eastern Pacific and decreased in the central and western Pacific during the warm phase, and a reversed

situation occurs during the cold phase. The magnitude of the thermocline depth anomalies, however, is underestimated in the models, especially over the eastern Pacific during the warm phase. This may contribute to the small subsurface temperature anomalies in most models during the warm phase. By comparison, CCSM3 + NR has a pattern close to that observed because of an improvement in the meridional extension.

To examine the subsurface signatures of ENSO asymmetry, we add the upper-ocean temperature anomalies over the two phases. We do the same sum for the depth of the thermocline. Figure 7 shows the results—the left panels show the asymmetry in the upper-ocean temperature anomalies and the right panels show the asymmetry in the thermocline depth anomalies. Observations indicate a positive asymmetry of subsurface temperature of  $1^{\circ}\text{C}$  around 75-m depth over the eastern Pacific, and a negative asymmetry of  $0.4^{\circ}\text{C}$  in subsurface temperature in the western Pacific. All of the models underestimate the asymmetry in the subsurface temperature. Therefore, the weaker SST asymmetry in the models is likely associated with an underestimate of the asymmetry in subsurface temperature. Again, CCSM3 + NR has an increased positive asymmetry in the subsurface temperature compared to other old versions (Fig. 7f). We have calculated the sum of warm and cold anomalies for the upper-ocean heat content (defined as the ocean temperature averaged over the upper 260 m). The asymmetry pattern of the ocean heat content is very similar to that of SST shown in the left panel of Fig. 3 (not shown). There is hardly any asymmetry of ocean heat content over the eastern Pacific in the four earlier versions, while the improvement is evident in CCSM3 + NR. The underestimate of the asymmetry in the subsurface temperature anomalies also shows up in the asymmetry in the depth of the thermocline (right panel of Fig. 7). The asymmetry in the observed thermocline depth anomalies is characterized by a broad negative region in the western Pacific. This feature is largely absent in the four earlier versions of the model. The latest version of the model has a negative asymmetry in the western Pacific, but the negative asymmetry is still too confined to the equator, and the maximum is located too far to the east.

Figure 8 further shows the meridional structures of the asymmetry in the upper-ocean temperature anomalies over the western Pacific and eastern Pacific. Observations are characterized by two maximum values of negative asymmetry off the equator in both hemispheres over the western Pacific (Fig. 8a) and a strong positive asymmetry right at the equator over the eastern Pacific (Fig. 8g). The negative asymmetry over the western Pacific covers a wider meridional region extending from  $10^{\circ}\text{S}$  to  $10^{\circ}\text{N}$  while the positive asymmetry over the

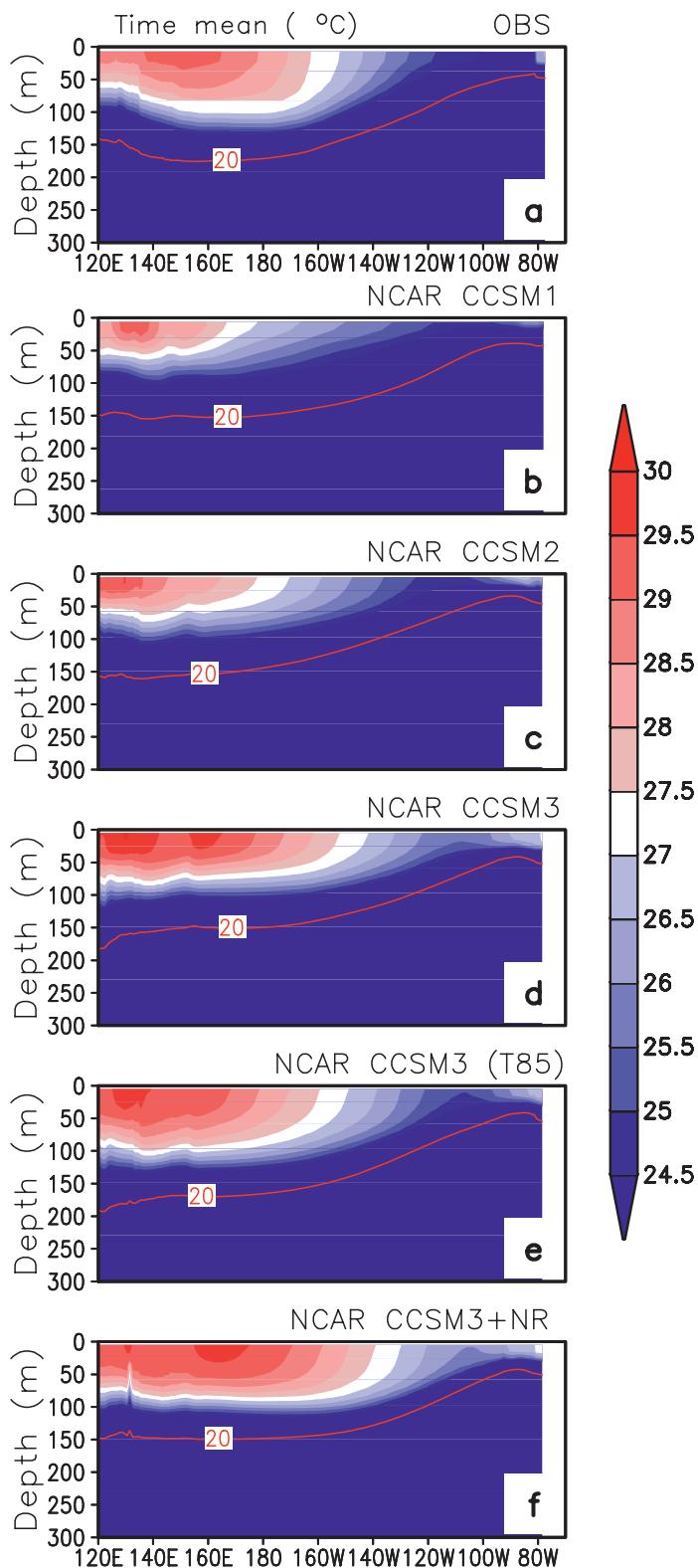


FIG. 5. Time-mean equatorial (5°S–5°N) upper-ocean temperature from five models and SODA data. The 20°C isotherm is plotted (red solid line). The length of data used in the calculation is 50 yr for all the five models and SODA data (1950–99).

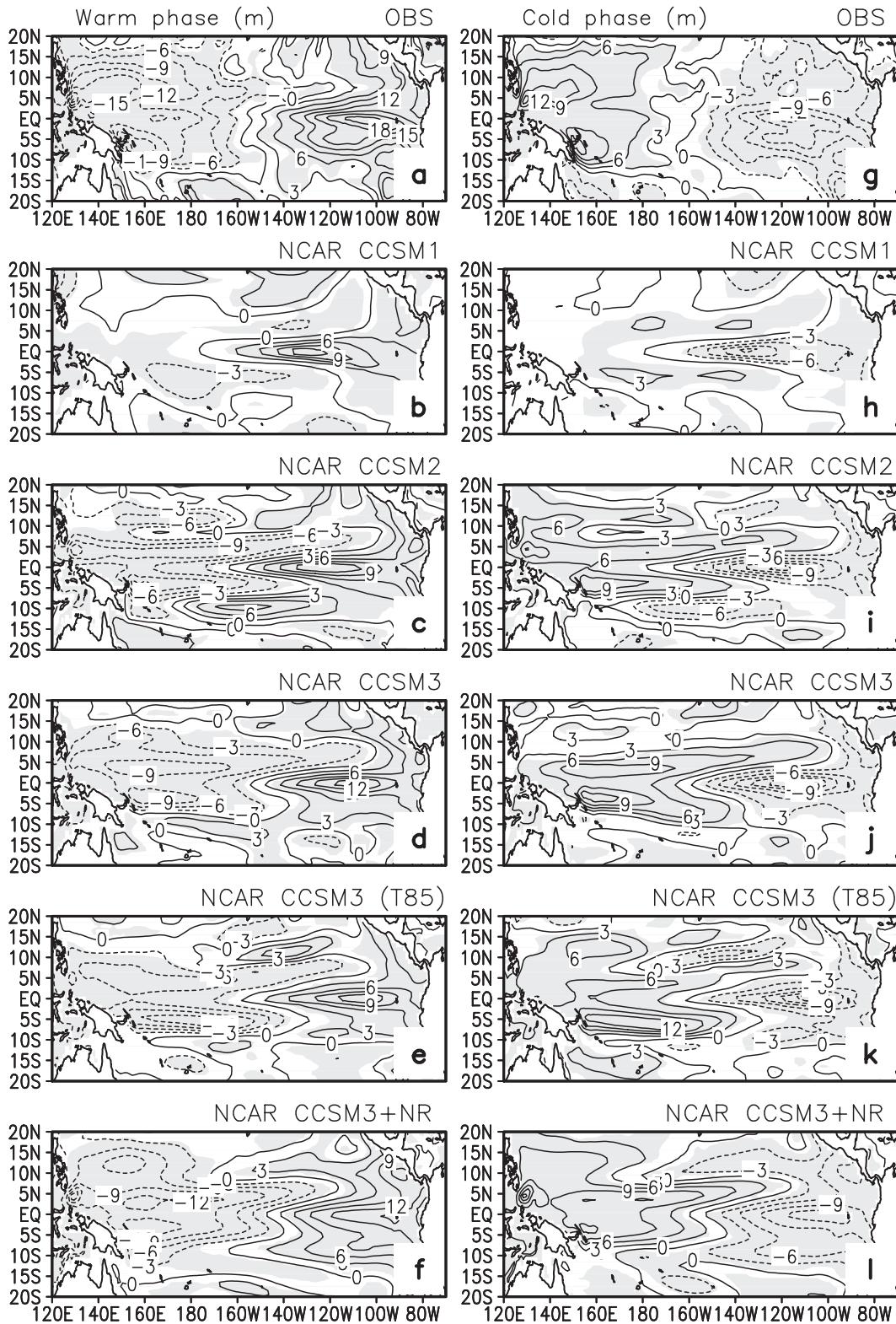


FIG. 6. Spatial distributions of 20°C isotherm depth anomalies during (left) warm and (right) cold periods. The length of data used in the calculation is 50 yr for all the five models and SODA data (1950–99). Shading indicates that the anomalies are statistically significant at the 95% confidence level using the  $t$  test.

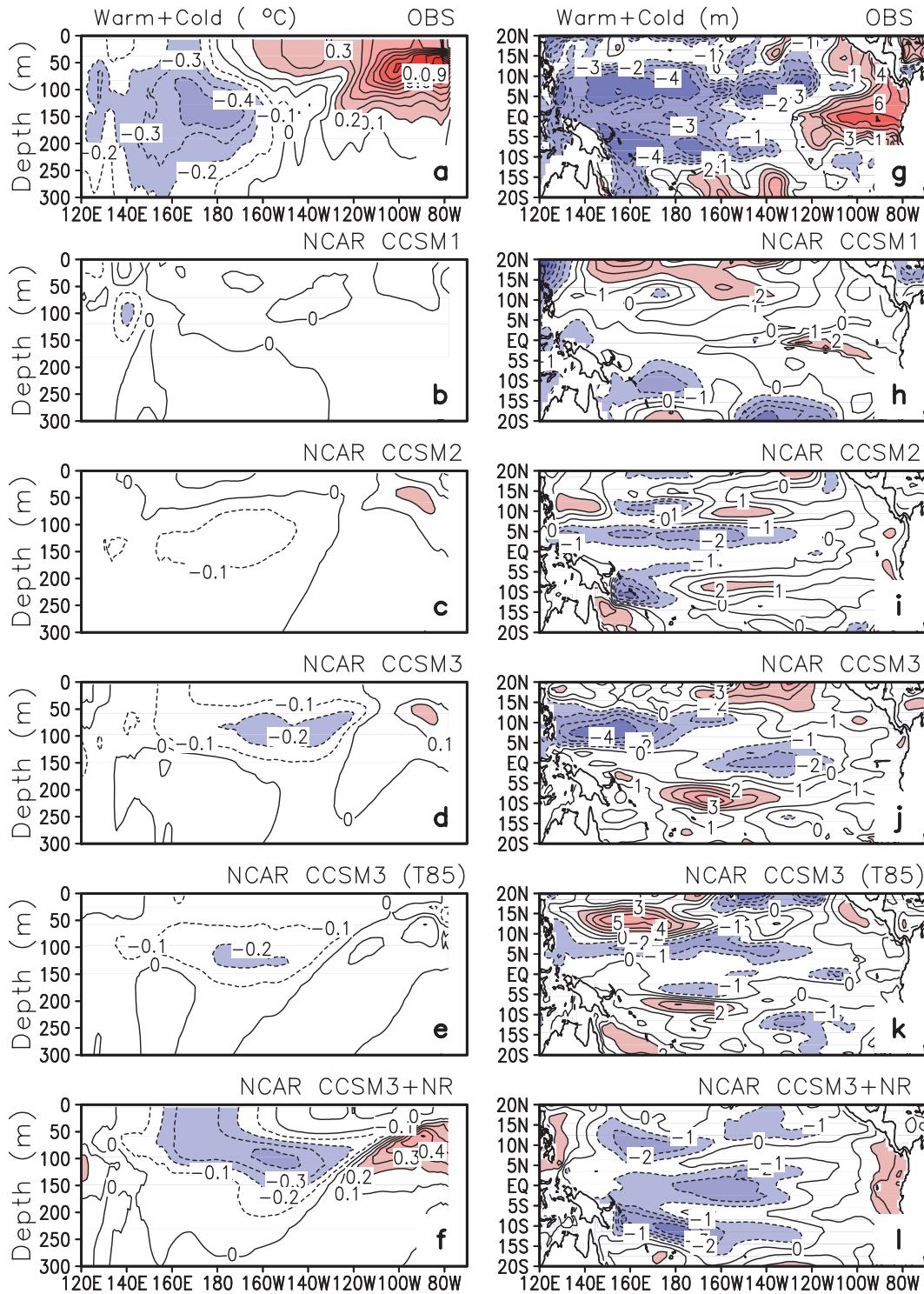


FIG. 7. The asymmetry in the equatorial upper-ocean (left) temperature and (right) 20°C isotherm depth. Shown are the results from the sum between warm anomalies and cold anomalies. The length of data used in the calculation is 50 yr for all the five models and SODA data (1950–99).

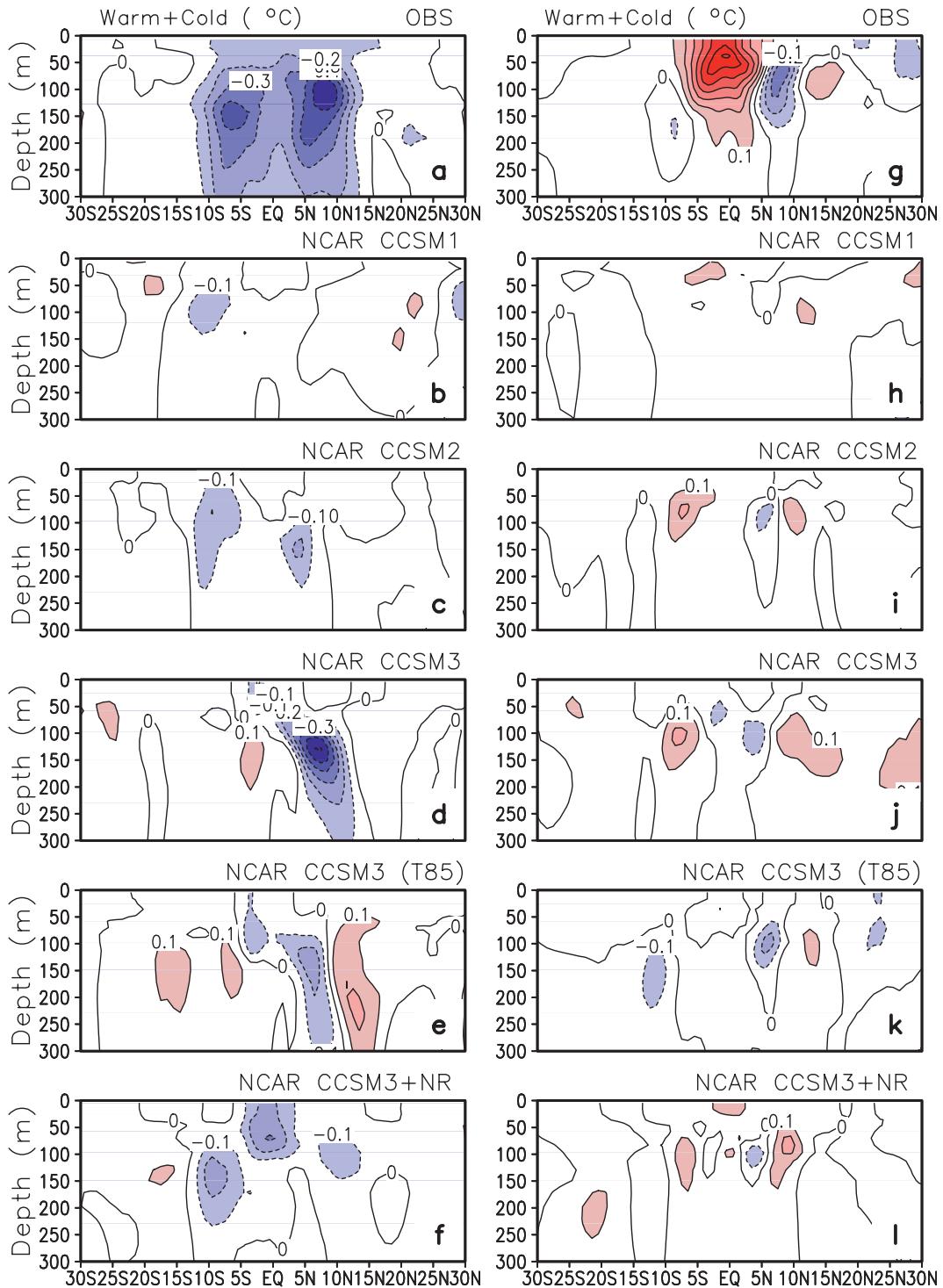


FIG. 8. The asymmetry in the meridional cross sections of the upper-ocean temperature over the (left) western Pacific ( $130^{\circ}$ – $170^{\circ}$ E) and (right) Niño-3 region ( $210^{\circ}$ – $270^{\circ}$ E). Shown are the results from the sum between the warm and cold anomalies. The length of data used in the calculation is 50 yr for all the five models and SODA data (1950–99).

eastern Pacific is mainly confined within the equatorial region ( $5^{\circ}\text{S}$ – $5^{\circ}\text{N}$ ). These negative and positive anomalies appear to be associated with the equatorial Rossby and Kelvin waves, respectively. Except for a better simulation of the negative asymmetry over the region of  $5^{\circ}$ – $10^{\circ}\text{N}$  in the western Pacific in CCSM3 (Fig. 8d), all of the models underestimate the negative asymmetry over the western Pacific and the positive asymmetry over the eastern Pacific. The improvement of the asymmetry in the meridional sections over the eastern Pacific in the new version of CCSM3 + NR is not as obvious as that in the zonal cross sections noted earlier (Fig. 8l). This is expected because of the cancellation between the positive asymmetry over the far eastern Pacific and the overly negative asymmetry over the region of  $150^{\circ}$ – $120^{\circ}\text{W}$ , as shown in Fig. 7f.

To obtain a qualitative measure of the asymmetry in the subsurface temperature, we plot the histogram of the monthly-mean Niño-3 subsurface temperature anomalies (Fig. 9). The figure shows that the observed Niño-3 subsurface temperature anomalies are skewed to the positive values and the maximum positive anomalies can reach more than  $6^{\circ}\text{C}$ . The underestimate of the maximum positive anomalies in the models ranges from  $5^{\circ}\text{C}$  in CCSM1 to  $2^{\circ}\text{C}$  in CCSM3 + NR. Although the overall distribution in the models is close to normal distribution, the new version of CCSM3 + NR has a more similar shape to that observed among the five models.

The skewness and variance of the subsurface temperature anomalies are further shown in Table 2. The results confirm that all of the models underestimate the variability and skewness of Niño-3 subsurface temperature. Comparing Table 2 with Table 1 indicates that the interannual variability of the Niño-3 subsurface temperature is almost as twice strong as that of Niño-3 SST in observations. However, the contrast in the strength of the variability between the surface and the subsurface in the models is not as significant as that observed. In fact, the subsurface temperature variability is even weaker than the SST variability in two old versions (CCSM1 and CCSM2). Table 2 also shows that in the models, there is a good correspondence between the weaker variability of subsurface temperature and the underestimate of the skewness of subsurface temperature over the Niño-3 region. From Table 2, we find that over the western Pacific the observed negative asymmetry of subsurface temperature ( $T_{\text{sub\_WP}}$ ) is underestimated in all of the models, which is consistent with the results shown in the left panel of Fig. 7. Further considering the underestimate of the positive asymmetry in Niño-3 subsurface temperature ( $T_{\text{sub\_Niño3}}$ ), we are not surprised at the findings that the underestimate of the positive asymmetry in the difference of subsurface temperature

between the Niño-3 region and the western Pacific ( $T_{\text{sub\_diff}}$ ) is more severe in all of the models, accompanied with an enhanced underestimate of the variability of subsurface temperature in most models (the underestimate of  $T_{\text{sub\_diff}}$  variability in CCSM3 + NR is not as much as that of  $T_{\text{sub\_Niño3}}$  variability because of an increase of  $T_{\text{sub\_WP}}$  variability in this model). Stronger SST variability does not necessarily result in a stronger SST skewness. For example, we note that CCSM2 and CCSM3 + NR have a larger SST variability than that of the observations but have a weaker skewness of SST. CCSM1 has a much weaker SST variability than CCSM2, but the skewness value in the former is much larger than that in the latter (Table 1). However, in the subsurface (Table 2), the models with smaller variance of subsurface temperature have a smaller skewness value and those with larger variance have a relatively larger skewness value over the Niño-3 region. It is thus suggested that the subsurface temperature anomalies may be a good proxy as a measure of the relationship between ENSO variability and asymmetry, because there seems to be no correlation between SST variance and SST skewness. Clearly, the present findings reveal that the lack of the asymmetry in SST field in the NCAR models is linked to an underestimate of the asymmetry in subsurface temperature.

### c. Possible causes for the weak ENSO asymmetry in the models

#### 1) ROLE OF SURFACE HEAT FLUX

Figure 10 shows the composite of the net surface heat fluxes for the two phases of ENSO. The figures show that four earlier versions of the NCAR model underestimate the negative heat flux anomalies during the warm phase. These models also underestimate the positive anomalies during the cold phase. The underestimation of surface heat flux anomalies in the four earlier versions is mainly caused by an underestimation of the negative anomalies of net surface shortwave flux during the warm phase and an underestimation of the positive anomalies of net surface shortwave flux during the cold phase. The contribution from the bias in the latent heat flux anomalies is relatively small (not shown). Regression of the surface heat flux onto the SST further reveals that the four earlier versions of the model underestimate the damping effect of the net surface heat flux on the local SST—the slope of the surface heat flux against the SST in these models is much weaker than that in the observations, especially for the slope against positive SST anomalies (Fig. 11). The difference between the slope against the positive SST anomalies and that against negative SST anomalies in the four earlier versions is also much

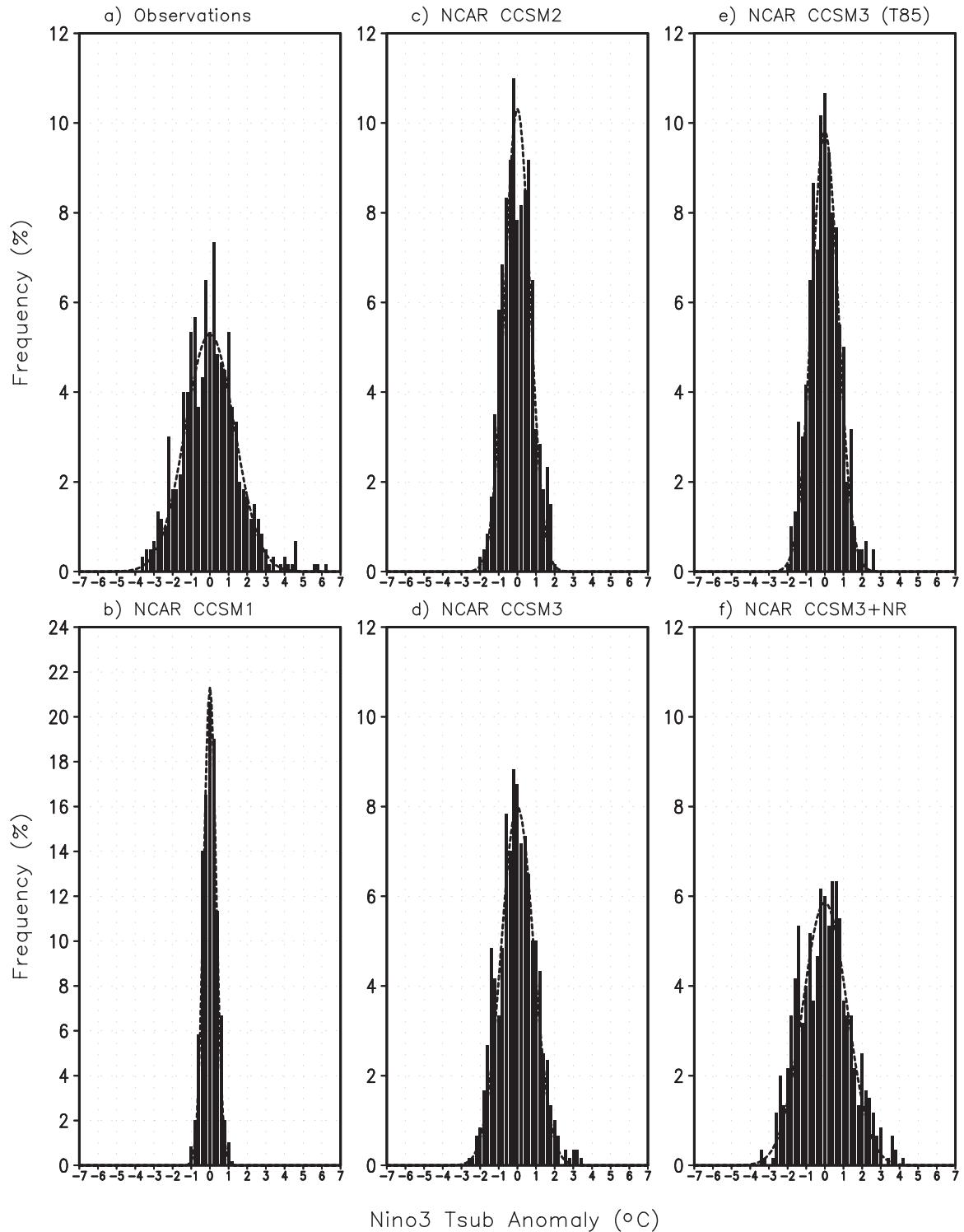


FIG. 9. Histograms of the monthly-mean Niño-3 subsurface temperature anomalies at 75-m depth. The normal distributions with the respective standard deviation are plotted (dashed line). A bin width of  $0.2^{\circ}\text{C}$  is used. The length of data used in the calculation is 50 yr for all the models and SODA data (1950–99).

TABLE 2. Statistics estimated for monthly time series of equatorial subsurface temperature anomalies from SODA data and five NCAR coupled models. The subsurface temperature anomalies at 75-m depth over the Niño-3 ( $5^{\circ}\text{S}$ – $5^{\circ}\text{N}$ ,  $210^{\circ}$ – $270^{\circ}\text{E}$ ) region (Tsub\_Niño3) and at 150-m depth over the western Pacific ( $5^{\circ}\text{S}$ – $5^{\circ}\text{N}$ ,  $130^{\circ}$ – $170^{\circ}\text{E}$ ) (Tsub\_WP), and from the difference between them (Tsub\_diff), are used in the calculation. The length of data for computing the statistics is 50 yr for all the five models and SODA data (1950–99).

Source	Standard deviation ( $^{\circ}\text{C}$ )			Skewness		
	Tsub_WP	Tsub_Niño3	Tsub_diff	Tsub_WP	Tsub_Niño3	Tsub_diff
Observed	1.36	1.51	2.44	−0.18	0.45	0.61
CCSM1	0.36	0.37	0.53	0.038	0.063	−0.065
CCSM2	1.21	0.77	1.53	0.016	0.15	0.089
CCSM3	1.10	1.00	1.66	0.092	0.23	−0.22
CCSM3 (T85)	1.08	0.81	1.54	0.10	0.21	0.010
CCSM3 + NR	1.42	1.36	2.36	0.22	0.24	0.11

smaller than that in the observations, indicating a less nonlinear relationship between surface heat flux and SST in these models. The surface heat flux–SST relationship is much improved in CCSM3 + NR ( $-18.4 \text{ W m}^{-2} \text{ K}^{-1}$  in the model versus the  $-21.5 \text{ W m}^{-2} \text{ K}^{-1}$  observed for positive SST anomalies, and  $-9.3 \text{ W m}^{-2} \text{ K}^{-1}$  in the model versus the  $-12.9 \text{ W m}^{-2} \text{ K}^{-1}$  observed for negative SST anomalies). The increase of net surface shortwave flux anomalies during two phases of ENSO is responsible for the improvement of surface heat flux anomalies in the new version. The new model also has an improved asymmetry pattern in the surface heat flux, although the magnitude of asymmetry is underestimated over the central and eastern Pacific (by  $\sim 10$ – $15 \text{ W m}^{-2}$ ; see Fig. 11) because of an overestimate of positive anomalies of net surface shortwave flux during the cold phase. The match of the stronger positive asymmetry of heat flux over the western Pacific (Fig. 11) with the stronger negative asymmetry of SST over that region (Figs. 3f,l) indicates a good negative correlation between the asymmetry of surface heat flux and the asymmetry of SST in the new model. It is clear that all of the NCAR models underestimate the damping effect of surface heat flux on SST over the central and eastern Pacific (right panel of Fig. 11). Thus, the underestimate of the ENSO asymmetry is not likely due to the contribution from the surface heat flux.

## 2) ROLE OF SURFACE WINDS

The precipitation anomalies during the warm and cold phases are shown in Fig. 12. Observations show that the maximum of the precipitation anomalies during the warm phase is located at about  $180^{\circ}$ , with a magnitude of  $4 \text{ mm day}^{-1}$ , whereas the maximum of the precipitation anomalies during the cold phase shifts westward and centers at  $170^{\circ}\text{E}$ , with a magnitude of  $-3 \text{ mm day}^{-1}$ . Such a shift is less obvious in the models. A weaker response in the precipitation during the warm phase over the equatorial central and eastern Pacific is evident in the four old versions, but one can see the progressive

improvement in these models in the location of the center of the response as well as the magnitude of the response of the precipitation. The center of maximum precipitation in CCSM3 + NR is even more eastward than that in the observations. The precipitation anomalies during the warm phase of ENSO over the eastern Pacific in this latest version are more comparable to that in the observations, although the maximum precipitation response in this new model is still somewhat weaker than that in the observations. CCSM3 + NR also starts to get the negative anomalies in the western Pacific, a feature that is almost entirely missing in the four old versions in the western Pacific.

The negative precipitation anomalies during the cold phase in the models (CCSM1, CCSM2, and CCSM3 at T42) are generally weaker than in the observations in the central Pacific. In the central Pacific, the negative precipitation anomalies in CCSM3 + NR are comparable to the observations. The precipitation anomaly in the far eastern Pacific in this latest version is even stronger than that in the observations, consistent with the stronger negative SST anomalies over that region (Fig. 2l). A common bias in all of these models is the lack of the positive precipitation anomalies in the western Pacific. Another common bias is a double rainband in the eastern Pacific, which exists in the warm phase as well, particularly for CCSM2, CCSM3 (T42), and CCSM3 (T85).

The associated zonal wind stress anomalies during the warm and cold phases are shown in Fig. 13. Observations show that the center of the equatorial zonal wind anomalies is located at about  $170^{\circ}\text{W}$  during the warm phase and at  $180^{\circ}$  during the cold phase. The westward shift of the center of the zonal wind anomalies during the cold phase relative to the warm phase is consistent with the shift in the observed precipitation anomalies. During the warm phase, the positive zonal wind stress anomalies are more confined within  $5^{\circ}\text{S}$ – $5^{\circ}\text{N}$ , and the magnitude is underestimated in four old versions. However, the progressive improvement is again evident. In contrast to the

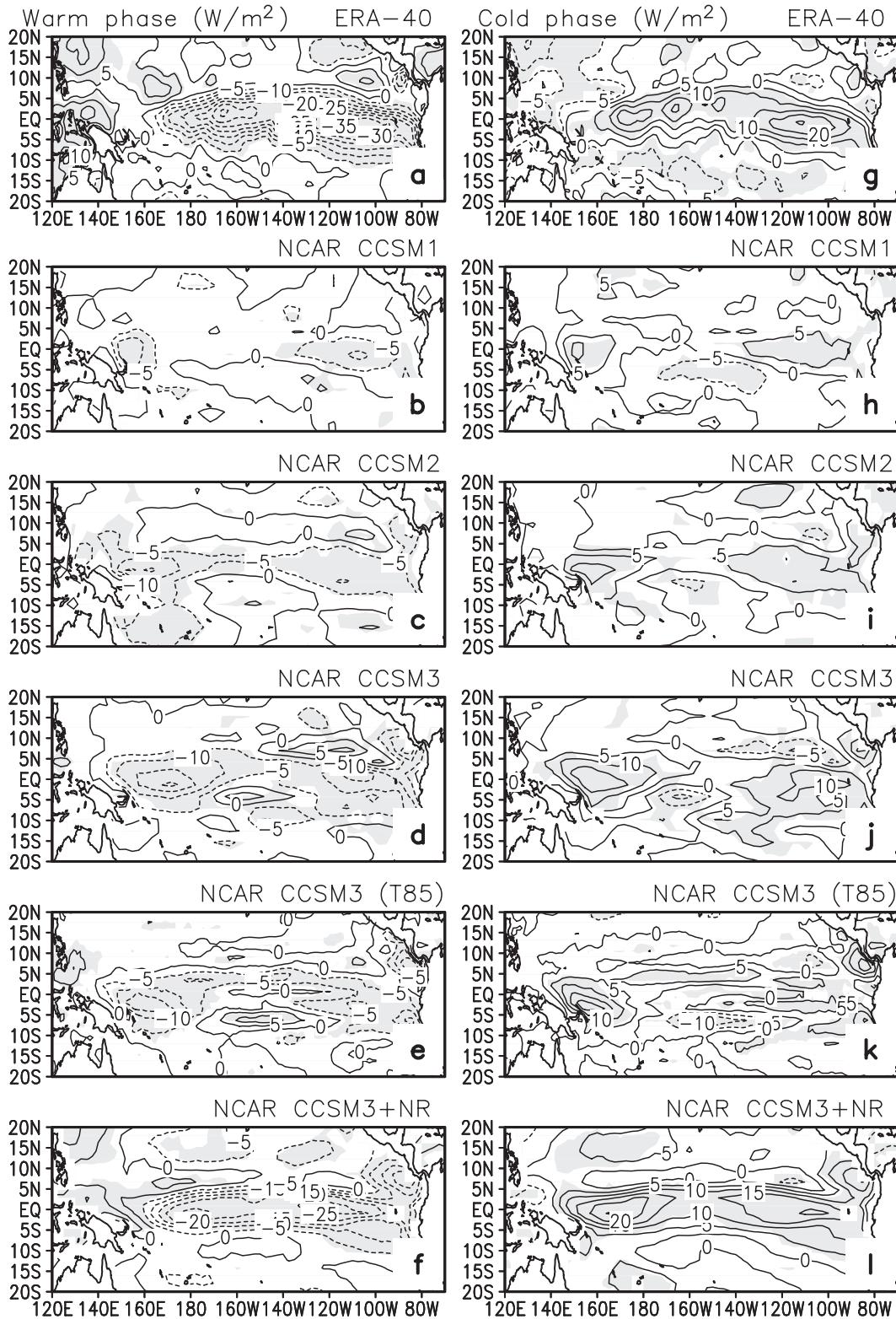


FIG. 10. Spatial distributions of net surface heat flux anomalies during (left) warm and (right) cold periods. The length of data used in the calculation is 50 yr for all the NCAR models and 45 yr for the ERA-40 (September 1957–August 2002). Shading indicates that the anomalies are statistically significant at the 95% confidence level using the  $t$  test.

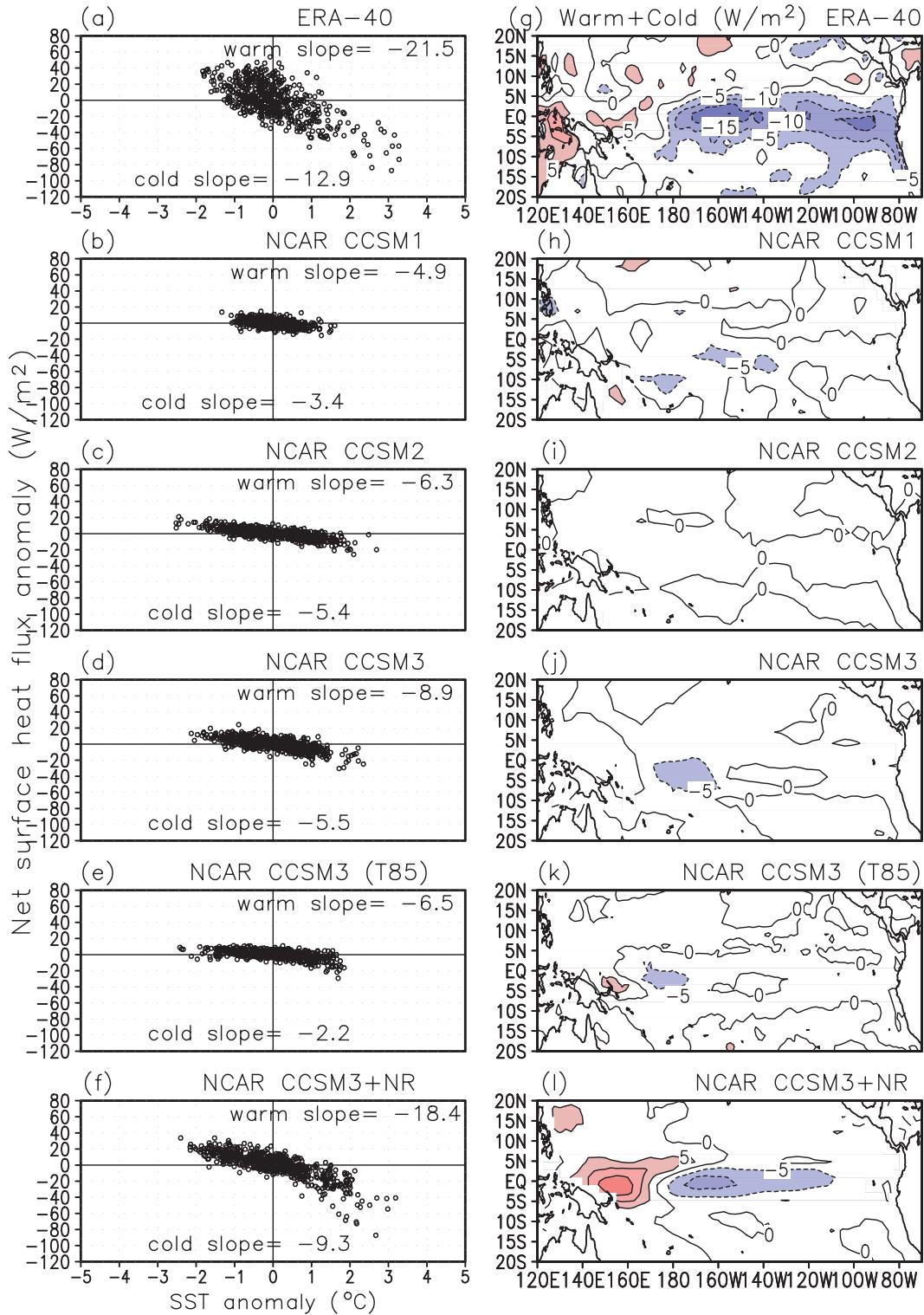


FIG. 11. (left) Scatterplot between the net surface heat flux anomalies and the SST anomalies over the Niño-3 region and (right) the spatial distributions of asymmetry in net surface heat flux. The corresponding slope values of linear regression for (top) positive and (bottom) negative SST anomalies are listed. The asymmetry in net surface heat flux is defined as the sum between warm anomalies and cold anomalies. The length of data used in the calculation is 50 yr for all the models and 45 yr for ERA-40 (September 1957–August 2002).

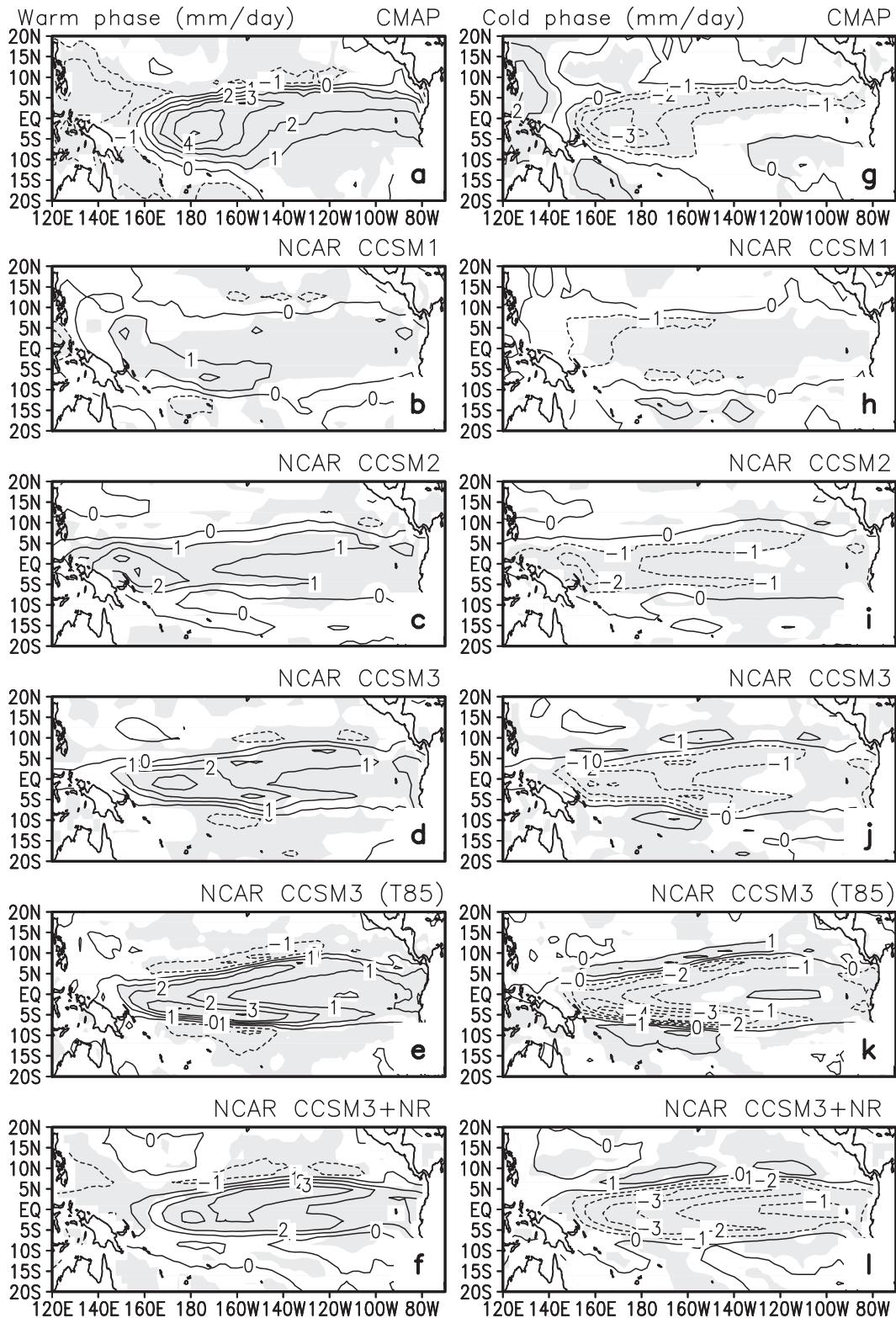


FIG. 12. Spatial distributions of precipitation anomalies during (left) warm and (right) cold periods. The length of data used in the calculation is 50 yr for all the NCAR models and 28 yr for the CMAP precipitation (1979–2006). Shading indicates that the anomalies are statistically significant at the 95% confidence level using the  $t$  test.

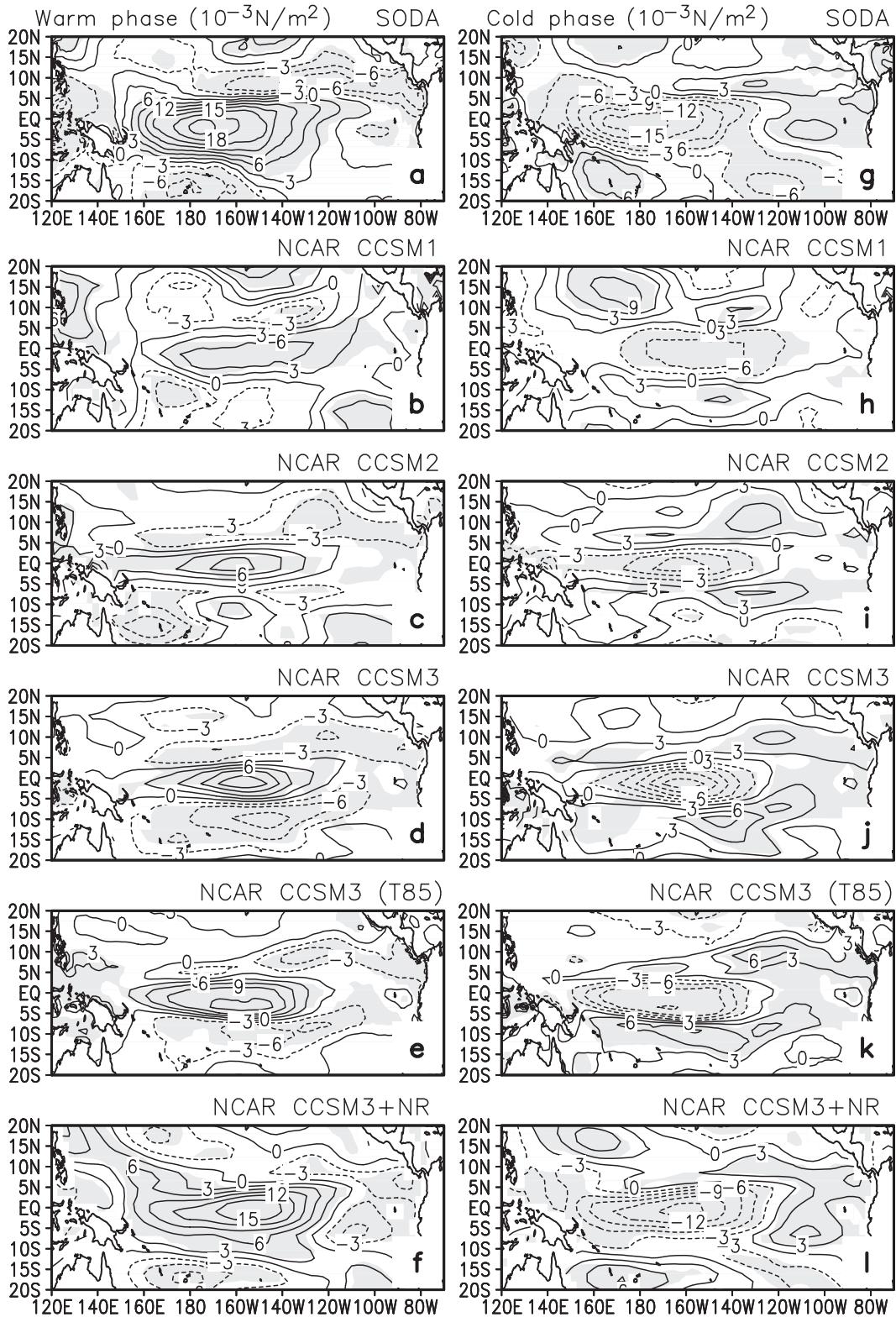


FIG. 13. Spatial distributions of zonal wind stress anomalies during (left) warm and (right) cold periods. The length of data used in the calculation is 50 yr for all the NCAR models and about 48 yr for SODA zonal wind stress (January 1958–November 2005). Shading indicates that the anomalies are statistically significant at the 95% confidence level using the *t* test.

four earlier versions, CCSM3 + NR is distinctly closer to the observations. It has an improvement not only in simulating the meridional extension of the zonal wind stress anomalies, but also in simulating the magnitude of the zonal wind stress anomalies. However, the center of the maximum wind stress anomaly is shifted eastward by about  $10^{\circ}$ – $20^{\circ}$  in all of the models during the warm phase (left panel of Fig. 13). During the cold phase the negative anomalies of the zonal wind stress are still more confined to the equatorial region of  $5^{\circ}\text{S}$ – $5^{\circ}\text{N}$  in the old models. CCSM3 + NR better captures the pattern of negative anomalies (right panel of Fig. 13). Again, the eastward shift of the center of the zonal wind stress anomalies is evident in all of the models relative to the observations during the cold phase. Compared to the four old versions, CCSM3 + NR has a wider zonal extension of stronger negative wind stress anomalies, ranging from  $170^{\circ}\text{E}$  to  $140^{\circ}\text{W}$  (Fig. 13l), which is associated with a wider zonal distribution of negative precipitation anomalies during the cold phase (Fig. 12l). The stronger negative zonal wind stress anomaly in CCSM3 + NR is also consistent with a stronger negative SST anomaly during the cold phase (Fig. 2l).

The sum between the warm composite anomalies and cold composite anomalies in precipitation and zonal wind stress is shown in Fig. 14. Observations show a strong asymmetry in precipitation fields: a strong positive asymmetry in the central and eastern Pacific and a strong negative asymmetry in the western Pacific (Fig. 14a). With the exception of CCSM3 + NR, such an asymmetry does not exist in the models. The underestimate of the asymmetry in the precipitation anomalies in the four old versions (Figs. 14b–e) is due to the lack of convection in the warm phase (Figs. 12b–e). An increase of precipitation in the warm phase over the central and eastern Pacific in the latest CCSM3 + NR improves the asymmetry of precipitation (Fig. 14f).

Consistent with the weak asymmetry in the precipitation anomalies, the asymmetry in the zonal wind stress anomalies is also weak in the NCAR models (right panel of Fig. 14). Again, the new version CCSM3 + NR has an improved asymmetry pattern of zonal wind stress anomalies because of the improvement in simulating the asymmetry in the precipitation anomalies (Figs. 14f,l).

The above analysis of the precipitation and zonal wind stress fields links the weak asymmetry in the latter in the models to the weak asymmetry in the former. Is the weak asymmetry in these two fields a consequence of the weak asymmetry in the corresponding SST fields or the cause of the latter? To answer this question, we have further analyzed the precipitation and zonal wind stress from the corresponding AMIP runs. Figure 15 gives the asymmetry pattern of precipitation and zonal wind

stress from the corresponding AMIP runs of the NCAR models, runs that are forced by the observed SST boundary conditions and therefore have identical SST patterns to those in observations. Although the SSTs now are the same as those observed, the asymmetry in precipitation over the eastern Pacific in the three old versions (CCM3, CAM2, and T42 CAM3) is notably weaker. The positive asymmetry in precipitation in T85 CAM3 and CAM3 + NR in the central and eastern equatorial Pacific is more comparable to the observations, but the corresponding negative asymmetry in the western Pacific ( $140^{\circ}$ – $170^{\circ}\text{E}$ ) is weaker than what is seen in the observations. In fact, all of the models underestimate the negative asymmetry in the precipitation response to ENSO forcing over the equatorial western Pacific ( $5^{\circ}\text{S}$ – $5^{\circ}\text{N}$ ,  $140^{\circ}$ – $170^{\circ}\text{E}$ ), and the underestimate of negative asymmetry ranges from  $1.0 \text{ mm day}^{-1}$  in T85 CAM3 to  $1.7 \text{ mm day}^{-1}$  in CAM2 (left panel of Fig. 15). This underestimate of the negative asymmetry over the western Pacific is mostly due to an underestimate of the drying during the cold phase in that region for all of the models except CCM3 (Fig. 16). The negative anomalies of precipitation in CCM3 over the equatorial western Pacific ( $140^{\circ}$ – $170^{\circ}\text{E}$ ,  $5^{\circ}\text{S}$ – $5^{\circ}\text{N}$ ) are close to that observed during the cold phase ( $-1.2 \text{ mm day}^{-1}$  in the model versus the  $-1.3 \text{ mm day}^{-1}$  observed), but during the warm phase the underestimate of negative anomalies is large ( $0.6 \text{ mm day}^{-1}$  in the model versus the  $-0.4 \text{ mm day}^{-1}$  observed) because of an overly westward extension of the positive anomalies in this model (Fig. 16b), resulting in the weaker negative asymmetry over the equatorial western Pacific shown in Fig. 15b. For the other four models, the underestimate of the negative anomalies over the equatorial western Pacific is small during the warm phase (ranging from  $0.3 \text{ mm day}^{-1}$  in CAM2 to  $0.4 \text{ mm day}^{-1}$  in CAM3). By comparison, the underestimate of the negative anomalies over the equatorial western Pacific is large during the cold phase (ranging from  $0.7 \text{ mm day}^{-1}$  in T85 CAM3 to  $1.4 \text{ mm day}^{-1}$  in CAM2). This bias is related to a lack of a significant westward shift of the location of the maximum response of precipitation from the warm phase to the cold phase. The response of precipitation to ENSO SST forcing in the models is apparently too local. This lack of spatial shift in the precipitation response is apparently also the cause of the positive weak asymmetry in the central and eastern Pacific (the models can capture the observed asymmetry in the magnitude response between the two phases: the maximum response during the warm phase is greater than the maximum response during the cold phase). With the exception of CAM3 + NR, the precipitation response in the eastern Pacific during the warm phase is too weak.

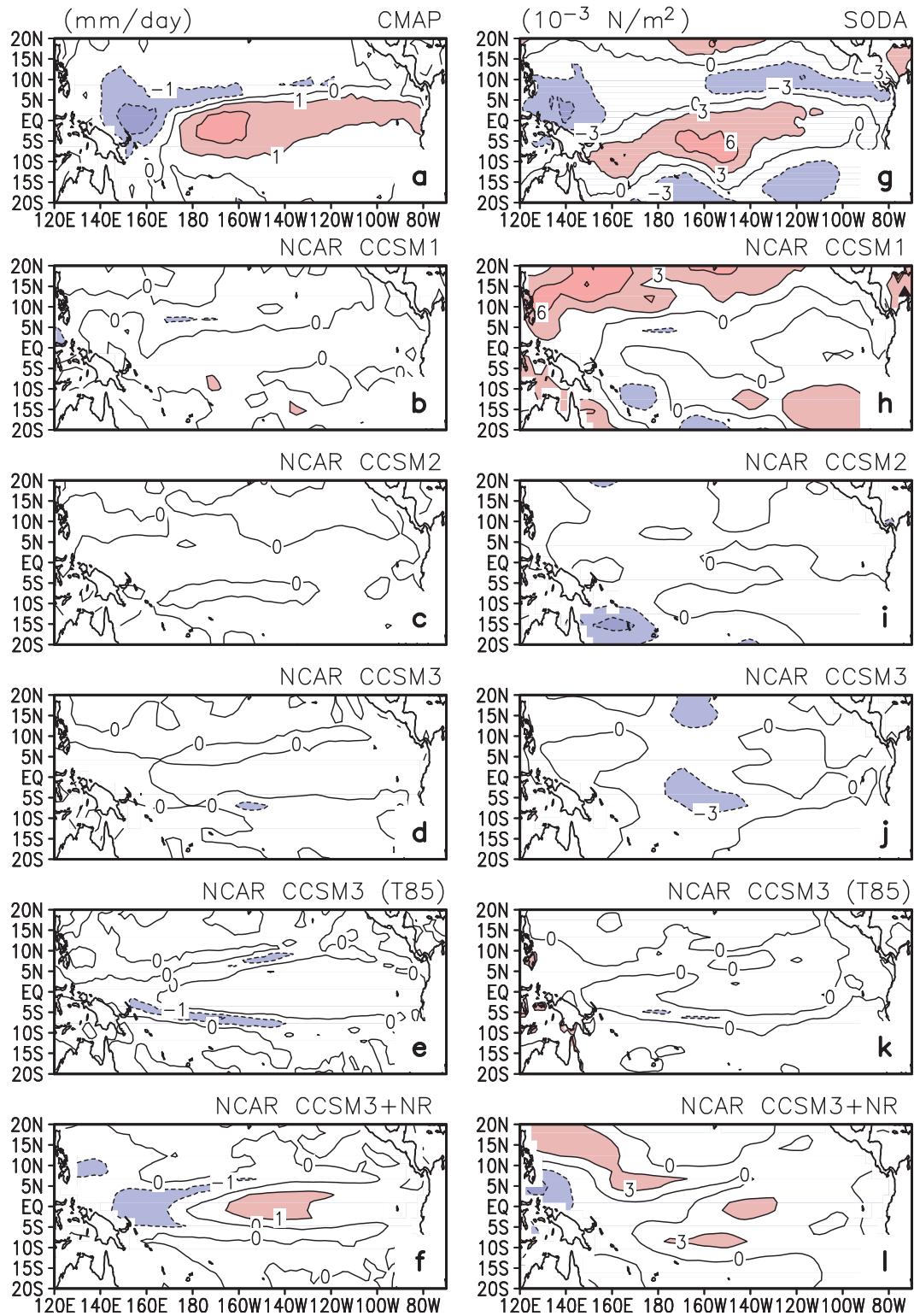


FIG. 14. Spatial distributions of asymmetry in (left) precipitation and (right) zonal wind stress. Shown are the results from the sum between warm anomalies and cold anomalies. The length of data used in the calculation is 50 yr for all the models and about 48 yr for SODA zonal wind stress (January 1958–November 2005), and 28 yr for the CMAP precipitation (1979–2006).

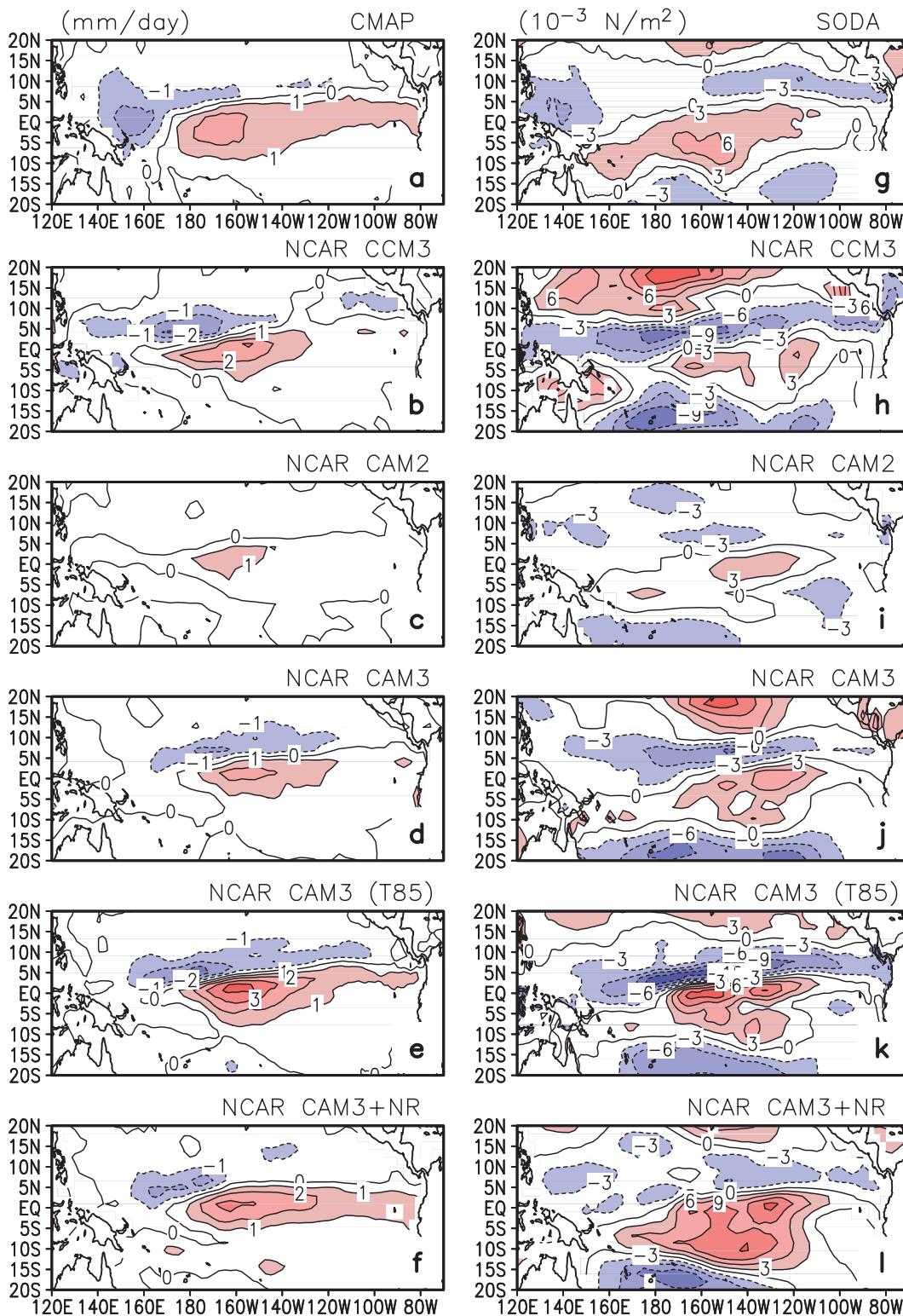


FIG. 15. As in Fig. 14, but for the corresponding AMIP runs. Shown are the results from the longest corresponding AMIP runs available to us, whose periods used in the calculation are February 1979–January 1993 for CCSM1, January 1950–December 1999 for CCSM2, January 1979–December 1999 for T42 CCSM3, January 1979–December 2000 for T85 CCSM3, and January 1977–December 2002 for CCSM3 + NR. The AMIP runs for the four old versions used in this study are the same as those used by Sun et al. (2009).

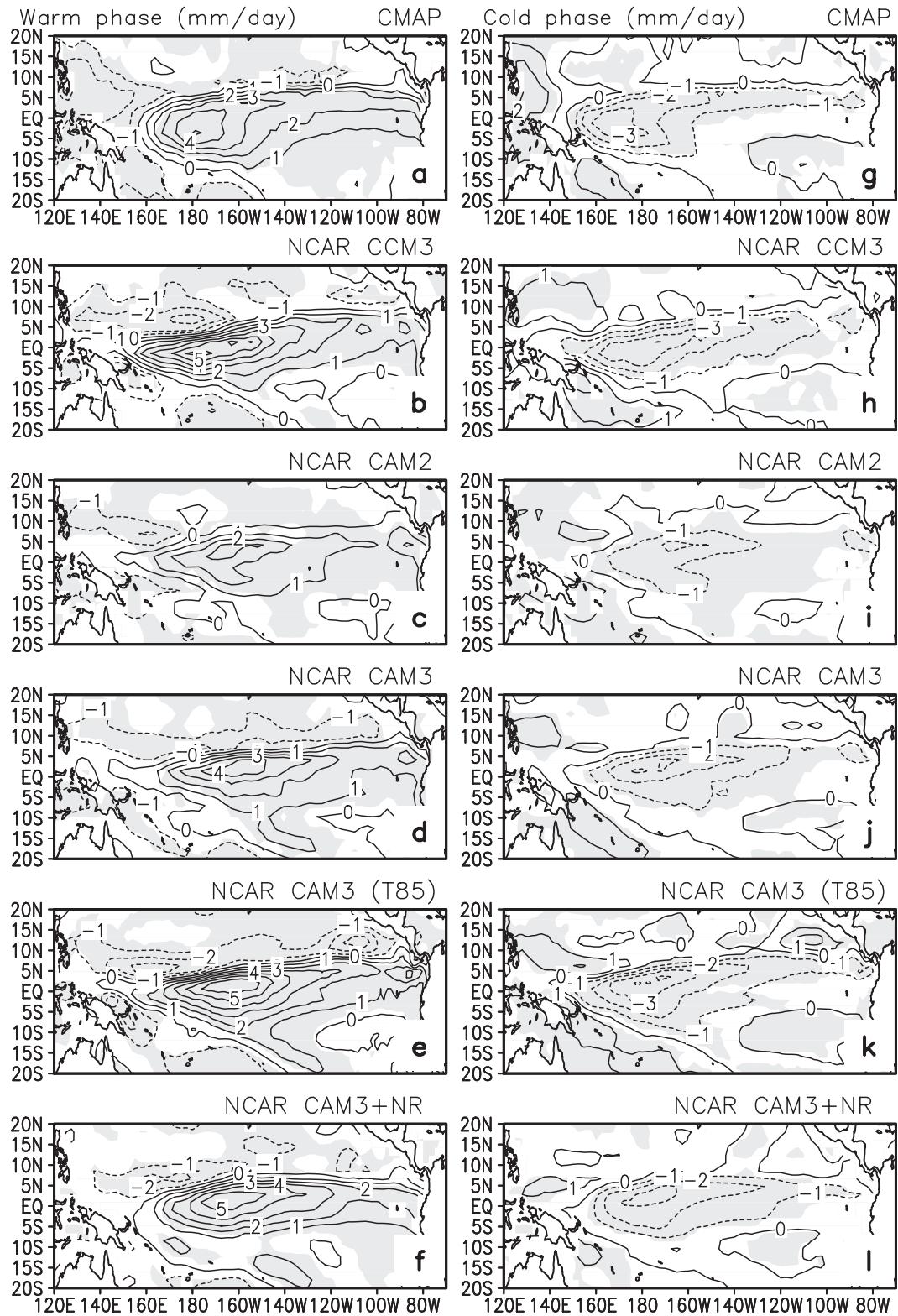


FIG. 16. As in Fig. 12, but for the corresponding AMIP runs. Shading indicates that the anomalies are statistically significant at the 95% confidence level using the *t* test.

Consistent with the lack of the westward shift in the pattern of precipitation response, there is also a lack of shift in the models in the pattern of zonal wind stress (Fig. 17). This bias is apparently the main cause of the substantially weak negative zonal wind stress asymmetry in the equatorial western Pacific in all of the models, and the weak positive zonal wind stress asymmetry in the central and eastern Pacific in CCM3, CAM2, and T42 CAM3 (see the right panel of Fig. 15). Despite the lack of the spatial shift between the two phases of ENSO, the positive asymmetry in the central and eastern equatorial Pacific in T85 CAM3 and in CAM3 + NR is more comparable to the observations. This is because the asymmetry in the magnitude response in these two models is actually greater than that in the observations. Measured by the maximum response in the central equatorial Pacific, the asymmetry in the magnitude in observations is about  $6 \times 10^{-3} \text{ N m}^{-2}$ . This asymmetry in maximum magnitude in the latest two models reaches more than  $9 \times 10^{-3} \text{ N m}^{-2}$  and  $9 \times 10^{-3} \text{ N m}^{-2}$ , respectively, primarily because of the much stronger zonal wind stress response in the warm phase (Fig. 17). A close examination of the difference between these two later models reveals that the zonal wind stress response during the warm phase extends more eastward in CAM3 + NR. Correspondingly, during the warm phase the negative response of the zonal wind stress over the far eastern equatorial Pacific in the latest version is also weaker, moving toward that observed (Fig. 17f). These differences are apparently linked to the difference in the precipitation response to the SST forcing in these two models, because the precipitation response in the eastern Pacific ( $150^{\circ}$ – $80^{\circ}$ W) is stronger in CAM3 + NR than in T85 CAM3, and therefore closer to that observed during the warm phase (Figs. 16a,e,f).

Overall, the results from the AMIP runs suggest that the weak asymmetry in the SST is likely due to a weaker asymmetry in the precipitation response to the SST forcing. In the observations, there is a significant westward shift in the maximum response in the precipitation from the warm phase to the cold phase. However, models do not simulate this shift. With the exception of the latest version, the precipitation response in the eastern Pacific during the warm phase also tends to be too weak. These biases are then reflected in the zonal wind stress, that is, with a weaker asymmetry in the zonal wind stress. Such a bias is then reinforced in the coupled models because the resulting weaker asymmetry in the zonal wind stress causes a weaker asymmetry in the SST through the subsurface dynamics (see Figs. 4 and 7). The latter then in turn generates an even weaker asymmetry in the precipitation, and thereby an even weaker asymmetry in the zonal wind stress.

#### 4. Summary

In this study, we have evaluated the ENSO asymmetry in five successive Community Climate System Model (CCSM) versions. We find that all of the versions underestimate the ENSO asymmetry. This underestimate is found in both the SST and the subsurface temperature signatures of ENSO. We also find that the net surface heat flux damps the ENSO asymmetry in the SST field and the models underestimate this damping effect. In light of the results from the corresponding AMIP runs, we suggest that the lack of asymmetry in the SST field of ENSO is due to the weaker asymmetry in the precipitation response in the models than in the observations to a given SST forcing. In particular, we attribute the weaker asymmetry to the lack of westward shift in the precipitation response from the warm phase to the cold phase in the models. The weaker asymmetry in the precipitation then causes a weaker asymmetry in the zonal wind stress response. In the process of coupling, this bias is then introduced to the subsurface through the wind forcing, and then to the SST response. The positive loop between the SST, precipitation, and zonal wind stress then reinforces the bias, causing a substantial underestimate of the asymmetry of ENSO in the coupled models.

It has long been known that the dependence of deep convection on the SST is nonlinear (e.g., Graham and Barnett 1987). The response of precipitation to a change in the sign of an anomalous SST forcing in the central and eastern Pacific is consequently asymmetric. (This nonlinearity is fundamentally related to the existence of the zonal SST gradient in the mean state.) An asymmetric response in the precipitation then causes an asymmetric response in the zonal wind stress (Hoerling et al. 1997; Kang and Kug 2002). Our study therefore underscores the importance of accurately simulating the nonlinear dependence of deep convection on the SST in climate models.

The evaluation also reveals that models with stronger variance in the Niño-3 SST than observations do not necessarily have a stronger SST skewness. By examining the nonlinearity in the Niño-3 SST time series from 24 coupled models participating in the El Niño Simulation Intercomparison Project (ENSIP), Hannachi et al. (2003) also found that some models have larger variance but have a small SST skewness value (see their Table 4). To the extent that the phase asymmetry is a fundamental feature of the observed ENSO, using the Niño-3 SST variance alone to measure the level of ENSO activity in the models can be misleading.

Among the five NCAR models, the latest (CCSM3 + NR) has the best simulation of the ENSO asymmetry,

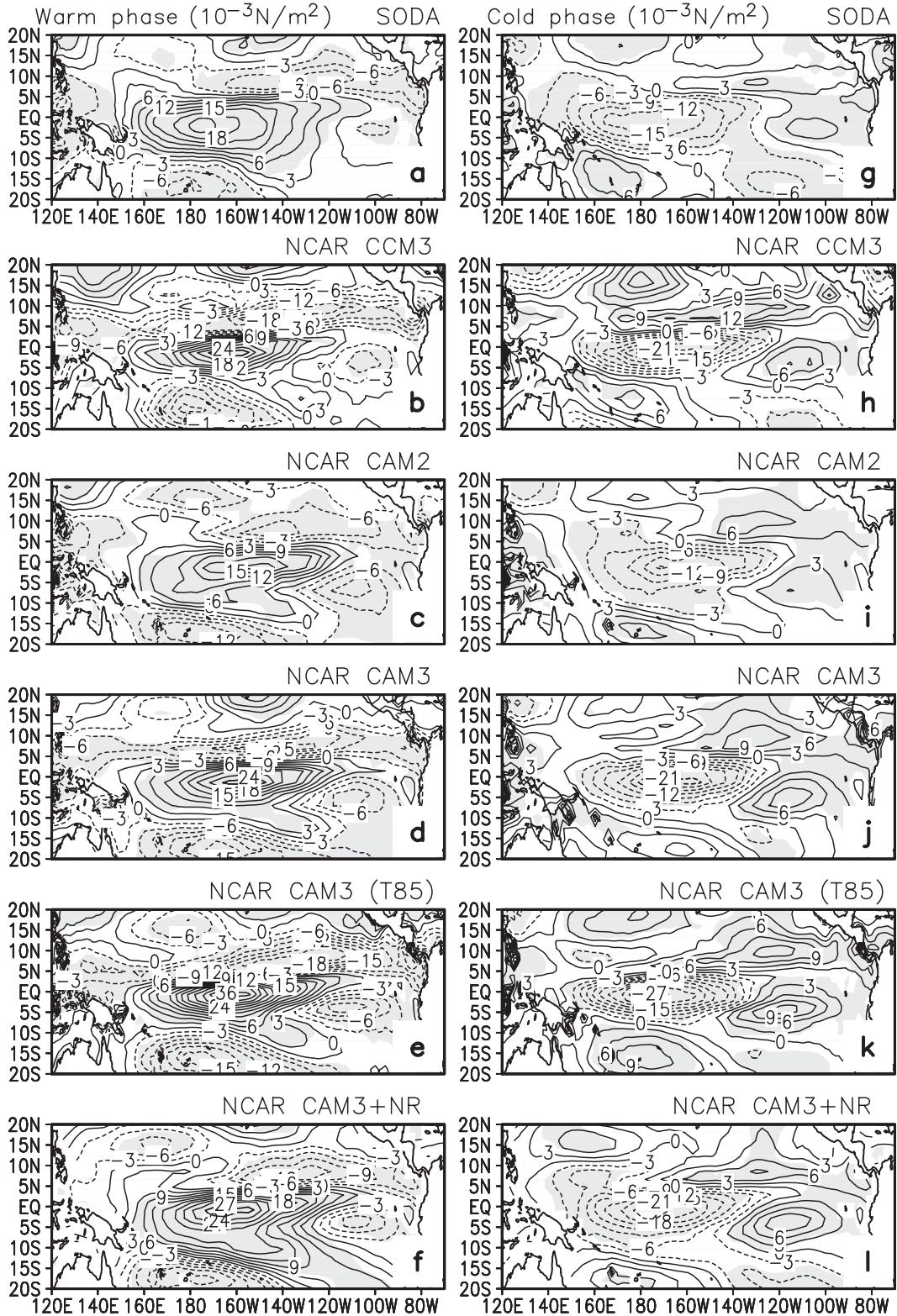


FIG. 17. As in Fig. 13, but for the corresponding AMIP runs. Shading indicates that the anomalies are statistically significant at the 95% confidence level using the *t* test.

although the asymmetry is still substantially weaker than that in the observations. This improvement appears to be a consequence of a stronger and broader response of precipitation (and zonal wind stress) to SST forcing during the warm phase of ENSO, which improves the asymmetry in the central Pacific in the response of precipitation to ENSO SST forcing and thereby in the response of zonal wind stress to ENSO SST forcing. The stronger asymmetry in the zonal wind stress then helps to create a stronger asymmetry in the subsurface temperature, which in turn contributes to a stronger asymmetry in the SST. Contrasting with earlier versions of the NCAR CCSM suggests that the increase in deep convection over the central and eastern Pacific during the warm phase in the latest CCSM3 + NR is a consequence of the replacement of Zhang and McFarlane deep convection scheme (Zhang and McFarlane 1995) with the Neale and Richter convection scheme (Neale et al. 2008).

The ENSO asymmetry over the eastern Pacific in CCSM3 + NR is still weaker than that in observations. In light of the results from the AMIP runs, the reason appears to be that although the positive response in the precipitation asymmetry over the central and eastern Pacific to ENSO forcing is adequately strong, a significant negative response in the precipitation asymmetry to ENSO forcing over the equatorial western Pacific is still missing because of the lack of a westward shift of convection during the cold phase. To improve the spatial shift in the location of deep convection in the models in response to ENSO, forcing appears to be a key step to take to further improve the simulation of ENSO, and its asymmetry in particular.

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